# Using SAGE for Search in Graph Theory 

Stephen G. Hartke

Department of Mathematics
University of Nebraska-Lincoln
www.math.unl.edu/~shartke2
hartke@unl.edu

Joint work with Jamie Radcliffe

## Searching in Graph Theory

Want to be able to do computations to
find or enumerate or classify graphs
(or graph-like structures) with a specified property.

## Moore Graphs

Def. A Moore graph is $\Delta$-regular, has $\Delta^{2}+1$ vertices, has diameter 2 , and is triangle- and $C_{4}$-free.

For which $\Delta$ do Moore graphs exist?

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Thm. (Hoffman-Singleton 60s)
Only for $\Delta=2,3,7$ and possibly 57.

## Graph Packing and Decomposition

Def. A packing of copies of $G$ in a graph $H$ is a set of edge-disjoint subgraphs $\left\{H_{1}, H_{2}, \ldots, H_{k}\right\}$ of $H$ such that $H_{i} \cong G$.

Def. A decomposition of $H$ into copies of $G$ is a packing of copies of $G$ such that every edge of $H$ appears in a copy.

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## Decomposition Problems

Conj. (Ringel) If $T$ is a tree with $r$ edges, then $K_{2 r+1}$ decomposes into copies of $T$.

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Conj. (El-Zanati) If $G$ is non-complete and has $r$ edges, then $K_{2 r+1}$ decomposes into copies of $G$.

## Petersen Graph

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$P$ is a Moore graph:
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$P$ has 15 edges,
$K_{10}$ has 45 edges.
$\Rightarrow 3$ copies of $P$.


## Petersen Graph

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No. Elegant proof using eigenvalues by Fan and Schwenk.

## Hoffman-Singleton Graph

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HS is a Moore graph:
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HS has 175 edges,
$K_{50}$ has 1225 edges.
$\Rightarrow 7$ copies of HS .


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This is a finite problem. Can we solve it using computer search?

## Integer Programming

An integer program IP is an optimization problem of the form

$$
\begin{array}{rrl}
\max & c^{\top} x & \\
\text { subject to } & A x & \leq b \\
& x & \geq 0 \\
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\end{array}
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$$

Linear programming relaxation drops the integral restriction.
LPs can be solved in polynomial time.
In practice, the best general method for attacking exponentially-sized problems.

## Feasibility Testing

To search for a decomposition, we construct an IP such that feasible solutions correspond to the desired decomposition.

## Petersen Graph

$\forall$ edge $e$ and copy $i$, indicator variables $x_{e}^{i} \in\{0,1\}$ indicating whether edge $e$ is in copy $i$.

Constraints:
partition: $\quad \forall$ edge $e, \sum_{i} x_{e}^{i}=1$
regular: $\forall$ copy $i, \forall v, \quad \sum_{e \ni v} x_{e}^{i}=3$

## Petersen Graph

$\forall$ edge $e$ and copy $i$, indicator variables $x_{e}^{i} \in\{0,1\}$ indicating whether edge $e$ is in copy $i$.
$\forall$ vertices $u, v, w$, copy $i$, ind vars $y_{u, v \rightarrow w}^{i} \in\{0,1\}$ indicating whether $w$ is a common nbr of $u$ and $v$ in copy $i$.

Constraints:
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Moore graph: $\quad \forall i, \forall u, v, \quad x_{\{u, v\}}^{i}+\sum_{w \neq u, v} y_{u, v \rightarrow w}^{i}=1$

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These properties uniquely determine the Moore graphs.
Feasible solutions to the IP are the desired decompositions.

## Feasibility Testing

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Some variables are fixed to 0 or 1, and the LP relaxation is tested for feasibility.

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Problem! By just renaming vertices, get another solution.

> Presence of multiple sols by symmetry makes branch-and-bound slower!

Solution: Only check one rep from each equivalence class.

## Canonical Representative

We pick a canonical rep from each equivalence class.
We test whether each partial decomp is a canonical rep.
Discard those that are not.

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Determining the canonical rep is computationally expensive. (closely related to graph isomorphism)

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For the Petersen graph $P$ :

$$
\left[\begin{array}{llllllllll}
- & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
- & - & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 \\
- & - & - & 0 & 0 & 0 & 1 & 1 & 0 & 0 \\
- & - & - & - & 0 & 0 & 0 & 0 & 1 & 1 \\
- & - & - & - & - & 0 & 1 & 0 & 1 & 0 \\
- & - & - & - & - & - & 0 & 1 & 0 & 1 \\
- & - & - & - & - & - & - & 0 & 0 & 1 \\
- & - & - & - & - & - & - & - & 1 & 0 \\
- & - & - & - & - & - & - & - & - & 0 \\
- & - & - & - & - & - & - & - & - & -
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$$

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Use same definition for a graph packing.

$$
\left[\begin{array}{llllllllll}
- & 3 & 3 & 3 & 2 & 2 & 2 & 0 & 0 & 0 \\
- & - & 2 & 2 & 3 & 2 & 0 & 3 & 0 & 0 \\
- & - & - & 0 & 2 & 3 & 3 & 0 & 2 & 0 \\
- & - & - & - & 0 & 0 & 2 & 2 & 3 & 3 \\
- & - & - & - & - & 3 & 0 & 2 & 0 & 3 \\
- & - & - & - & - & - & 0 & 0 & 3 & 2 \\
- & - & - & - & - & - & - & 3 & 2 & 3 \\
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\end{array}\right]
$$

## Canonical Rep Functions

Read and Faradžev: Represent partial packing as a matrix.
Define canonical rep as the max under lex order by rows.
Better: order by copy 2 first, then copy 1.
$\left[\begin{array}{llllllllll}- & 2 & 2 & 2 & 0 & 0 & 0 & 0 & 0 & 0 \\ - & - & 0 & 0 & 2 & 2 & 0 & 0 & 0 & 0 \\ - & - & - & 0 & 0 & 0 & 2 & 2 & 0 & 0 \\ - & - & - & - & 0 & 0 & 0 & 0 & 2 & 2 \\ - & - & - & - & - & 0 & 2 & 0 & 2 & 0 \\ - & - & - & - & - & - & 0 & 2 & 0 & 2 \\ - & - & - & - & - & - & - & 0 & 0 & 2 \\ - & - & - & - & - & - & - & - & 2 & 0 \\ - & - & - & - & - & - & - & - & - & 0 \\ - & - & - & - & - & - & - & - & - & -\end{array}\right],\left[\begin{array}{llllllllll}- & 0 & 0 & 0 & 1 & 1 & 1 & 0 & 0 & 0 \\ - & - & 1 & 1 & 0 & 0 & 1 & 0 & 0 & 0 \\ - & - & - & 0 & 1 & 0 & 0 & 0 & 1 & 0 \\ - & - & - & - & 0 & 1 & 0 & 1 & 0 & 0 \\ - & - & - & - & - & 0 & 0 & 0 & 0 & 1 \\ - & - & - & - & - & - & 0 & 0 & 1 & 0 \\ - & - & - & - & - & - & - & 0 & 1 & 0 \\ - & - & - & - & - & - & - & - & 0 & 1 \\ - & - & - & - & - & - & - & - & - & 1 \\ - & - & - & - & - & - & - & - & - & -\end{array}\right]$
Allows us to fix the first copy.

## Implementation

Originally implemented in C using the COIN LP library.
COIN: COmputational INfrastructure for Operations Research
www.coin-or.org
Open-source (but CPL), sponsored by IBM and ppl at Lehigh OSI: Open Solver Interface, for calling LP and IP solvers

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Solution: Python and SAGE!

## Python and SAGE

## Pros:

- Python is a readable, expressive high-level language. Great for experimentation!
- SAGE unites many open programs in a coherent way.
- SAGE is free-cost and freedom.
- SAGE has its own useful math code (ie, NICE).
- SAGE is documented.
- PyCOIN Python interface to COIN generated by SWIG.


## Python and SAGE

## Cons:

- New language to learn.
- Missing from GAP interface: Stabilizer, CosetRep.
- PyCOIN has memory leaks.
- Python is slower than straight C.


## Computational Results

Reimplemented in Python and SAGE over Thanksgiving 2007.

Run on a Pentium IV 3.4GHz Linux PC.
For Petersen graph, IP has 1215 vars, 3124 constraints
Without canonical rep func, takes 30 mins and 776 nodes.
With canonical rep func, takes a few secs and 66 nodes.

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For Petersen graph, IP has 1215 vars, 3124 constraints
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With canonical rep func, takes a few secs and 66 nodes.

For Hoffman-Singleton graph, IP has 361374 vars, 2529618 constraints

Work focusing now on refining the implementation details.

## Other Tools?

McKay has a powerful canonical rep function in nauty. Implemented as NICE in SAGE by Miller.

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Use of cvxopt and GLPK?

## Future Work

Resolve the Hoffman-Singleton decomp of $K_{50}$
Use method for other decomposition problems:
Conj of El-Zanati: If $G$ is non-complete and has $r$ edges, then $K_{2 r+1}$ decomposes into copies of $G$.

Smallest unknown case: $G=K_{6}-e$.

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Smallest unknown case: $G=K_{6}-e$.
Use in proofs?

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