

# A Comparison Between Two Matching Heuristics

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# What was implemented

**Algorithm:** Edmonds Cardinality Matching Algorithm (The blossom algorithm) implemented according to Korte, Vygen 2002, pg. 224-225.

Runtime:  $O(|V|^3)$

## ***Heuristics:***

### **1. Greedy Maximal Matching.**

Runtime:  $O(|E|)$

### **2. Expand-Contract** after specification by Adrian (EC)

Runtime:  $O(|E||V|)$

# Edmonds Blossom Algorithm For Unweighted Graphs

- Works by growing an alternating forest, and augmenting vertex-root paths.
- When odd cycles occur, vertices in the alternating tree might be both even and odd at the same time.
- To manage the odd cycles, add some additional logic to shrink cycles into pseudo-vertices.
- Proceed as usual.

# Expand-Contract

1. **Expand:** Turn the graph into a Bipartite graph
  - a) For each vertex  $v$ , create  $v1$  and  $v2$
  - b) For edge  $(v, w)$  create  $(v1, w2)$  and  $(w1, v2)$
2. **Find a maximum matching in the bipartite graph.**
3. **Contract:** Turn the graph into the original graph again.  
For all  $(vx, wy)$  in  $M$ , let all  $(v, w)$  be the found “matching”  $M'$ .
4. **Repair:** Look at the graph induced by  $M'$ 
  - a) Find all paths of even length and fix them.
  - b) Find all even cycles of even length and fix them.
  - c) Find all Odd cycles of even length and fix them.

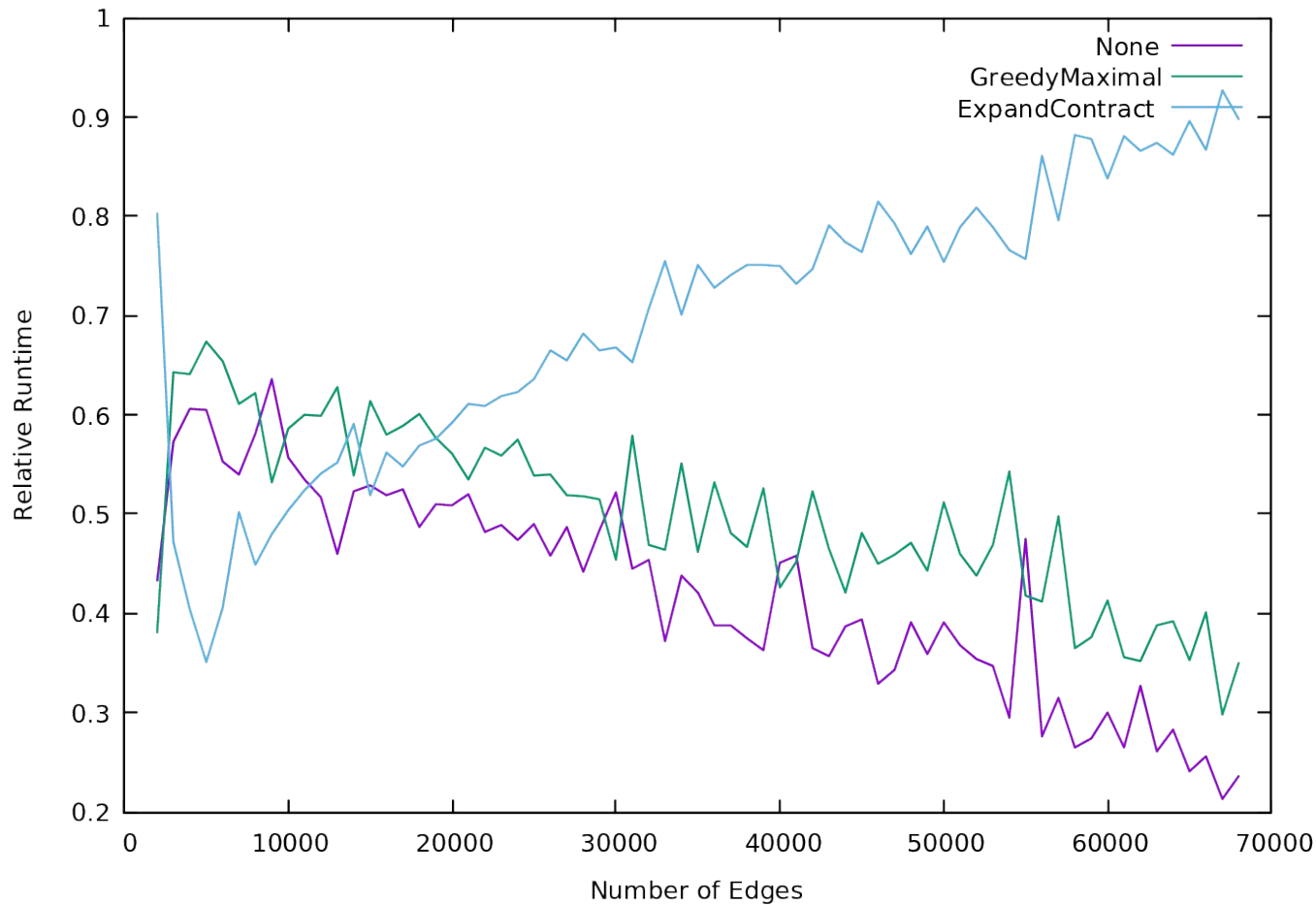
We fix a component by removing every second edge.

# Runtime Comparison Experiment

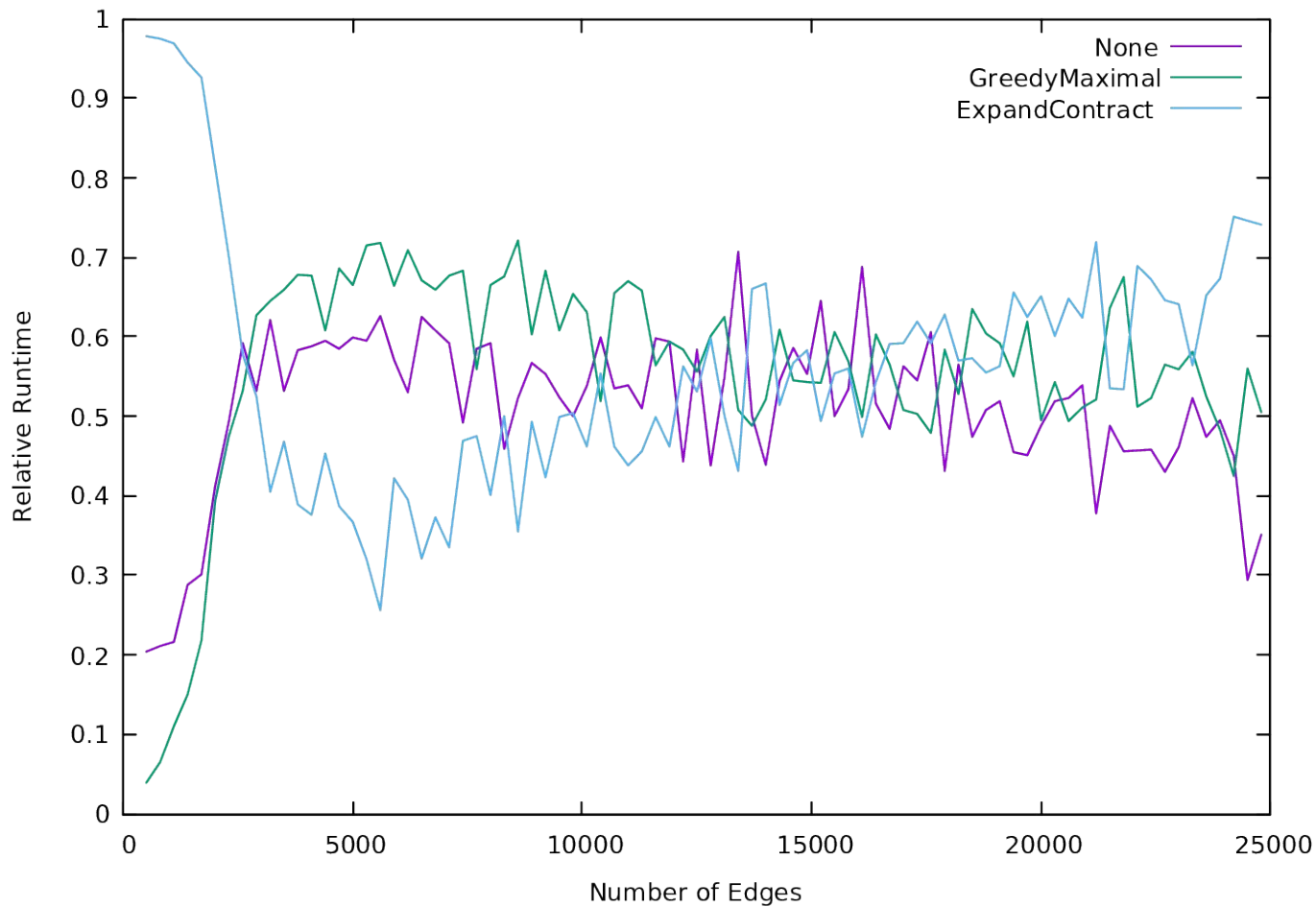
## Procedure:

1. Generate a random graph of a specific size.
2. Start timer.
3. Find a maximal matching with a specified Heuristic.
4. Run Edmonds Blossom algorithm with the maximal matching.
5. Stop timer.
6. Compare relative runtimes.

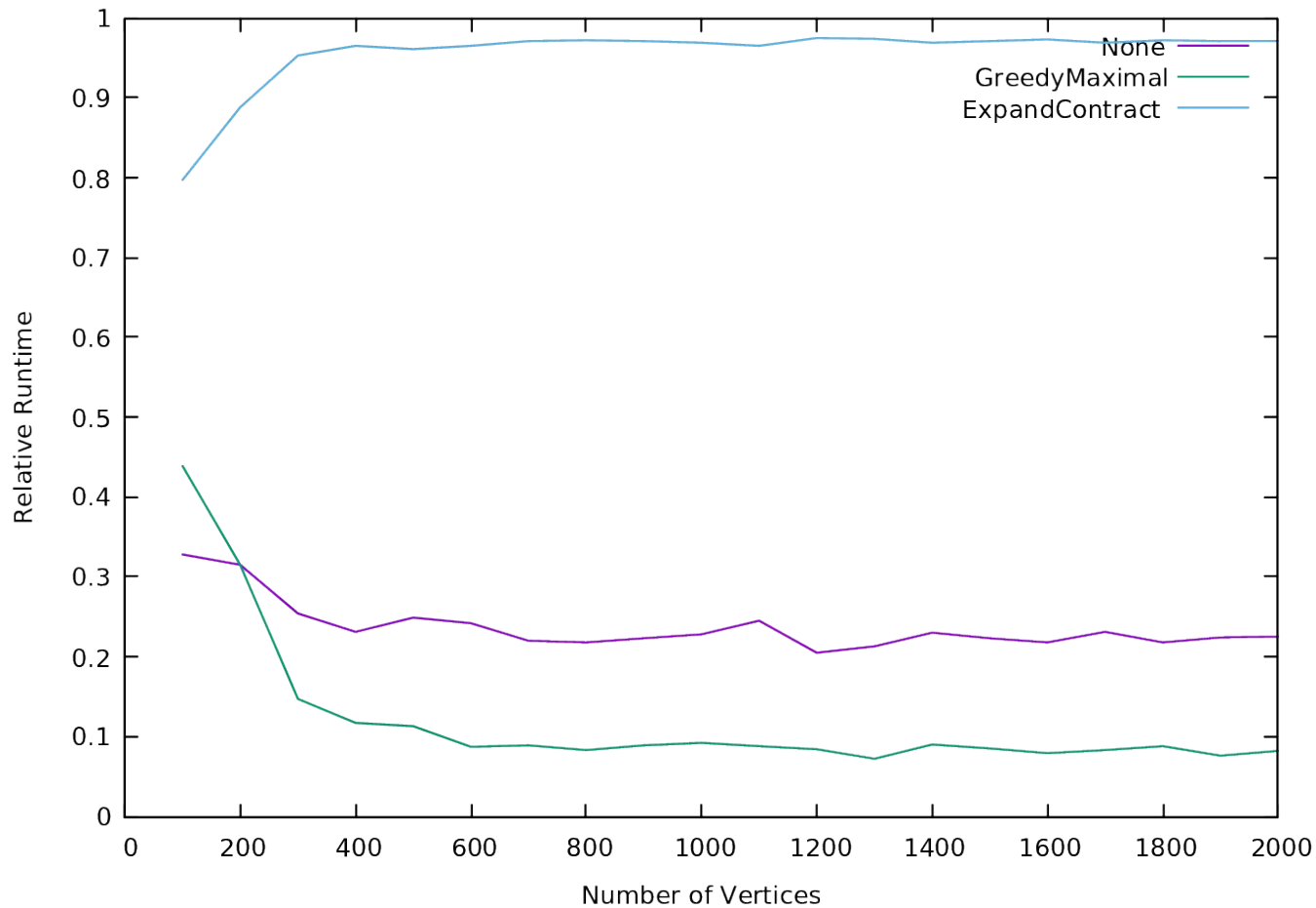
Relative Runtime: Vertices: 1000, Edges 1000-70000



Relative Runtime. Vertices: 1000, Edges 100-250000



Relative Runtime. Vertices: 100-2000, Edges: Same as vertices





# Experiment Conclusions: Runtime

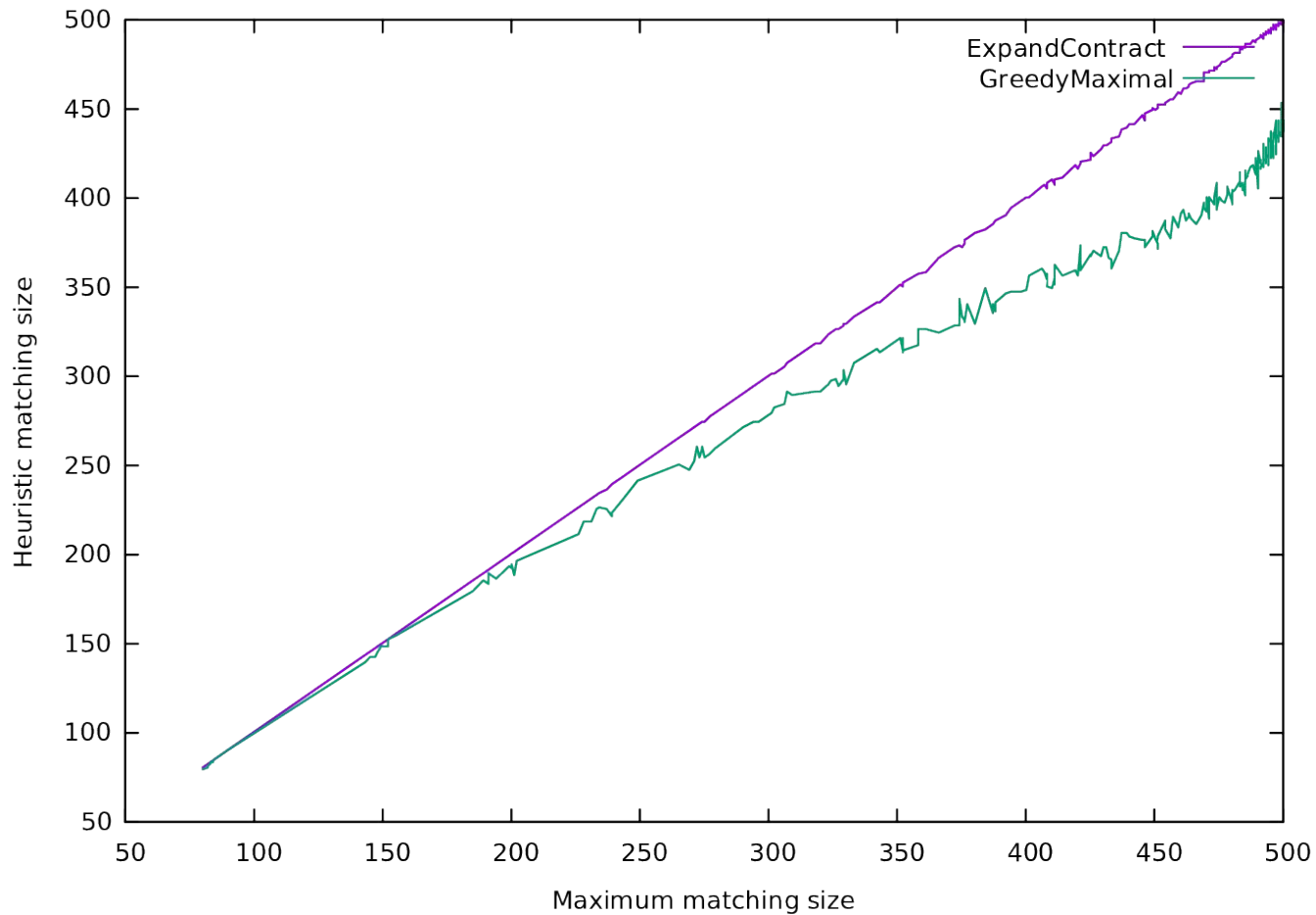
1. Finding a Greedy Maximal matching first did not influence the runtime significantly.
2. Expand-Contract is fast for graphs of a specific vertex-edge ratio.
3. Expand-Contract is very slow for very dense graphs.
4. Changing the graph size did not change the relative runtimes of the Heuristics when keeping  $|\text{vertex}| : |\text{edge}|$  ratio fixed.

# Matching Size Experiment

Procedure:

1. Generate a random graph.
2. Find a matching with Greedy Maximal.
3. Find a matching with Expand-Contract.
4. Compare the two matching sizes. with Maximum matching.

Matching found: Maximum Matching size: 75-500



# Experiment Conclusions: Matching Size

1. For very sparse graphs, Greedy-Maximal and Expand-Contract was fairly identical.
2. As denseness increases, Greedy-Maximal fell off but Expand-Contract almost always finds a near-maximum matching.
3. The average matching found by Greedy-Maximal was a  $\sim 0.88$ -approximation of  $M^*$  while Expand-contract found a  $\sim 0.98$ -approximation of  $M^*$ .

# Programming language

## Haskell Pros:

- Good set of standard data-structures to work with.
- Very easy to define new data structures.
- Pure functional programming forces a cleaner approach, and to think in terms of Types and Data instead of memory manipulation.

## Haskell Cons:

- Random access requires non-idiomatic data-structures that makes programming difficult. You can not easily define a mutable array. Most Algorithm descriptions in textbooks assume imperative programming with mutable state.

# Difficulties and Shortcomings

## Difficulties:

- Understanding the specification of the algorithm.
- Translating the low level specification into a higher-level view suitable for implementation in haskell.

## Known improvements to the algorithm:

- Use constant time data-structures for accessing state. Current implementation is based on Integer Sets which has a  $\log(n)$  lookup but at most the number of bits of an integer.