

# Parallel Dynamic Tree Contraction

## via Self-Adjusting Computation

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<sup>1</sup>Carnegie Mellon University, USA

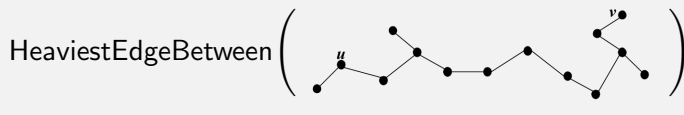
<sup>2</sup>Inria, France

<sup>3</sup>ITMO University, Russia

July 2017

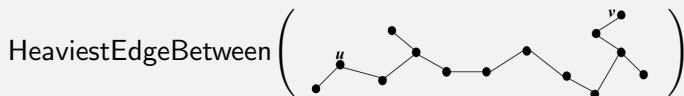
# Introduction

## Dynamic Algorithms

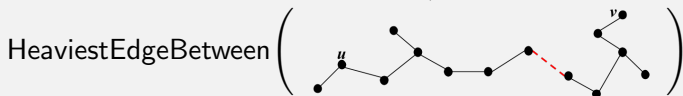


# Introduction

## Dynamic Algorithms

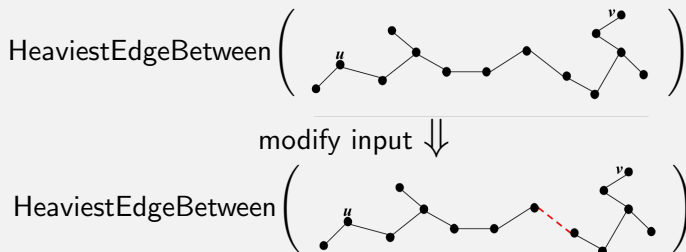


modify input  $\Downarrow$



# Introduction

## Dynamic Algorithms



## Efficiency?

- Lots of work on **sequential algorithms** with **unit changes**.
  - ▶ e.g., single edge insertion/deletion
- More efficient to apply **many changes** simultaneously.

# Contributions

## Question

In a forest of size  $n$ , how efficiently can we recompute some desired property after applying a batch of  $m$  changes (insertions/deletions of edges/vertices)?

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	Work	Span
Construction	$O(n)$	$O(\log^2(n))$
Update	$O\left(m \log \frac{n+m}{m}\right)$	$O(\log(n+m) \log(m))$

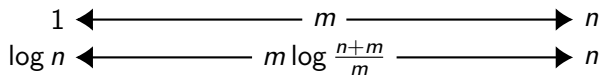
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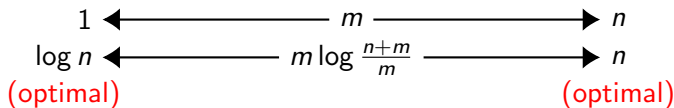
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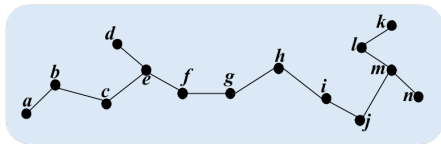
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- Miller, Reif (1985)
- *rake* and *compress*
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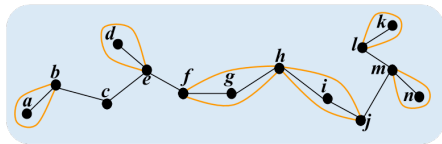
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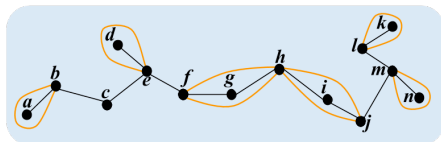
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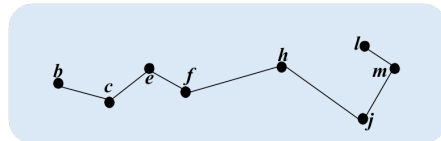
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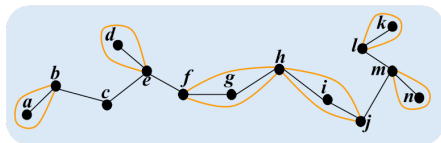
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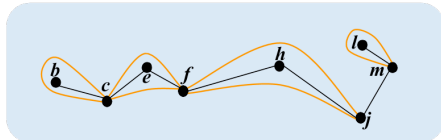
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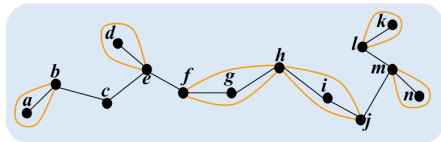
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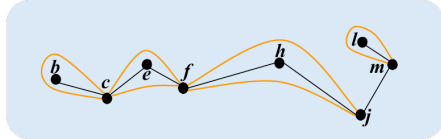
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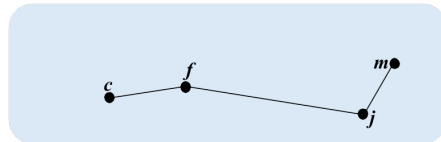
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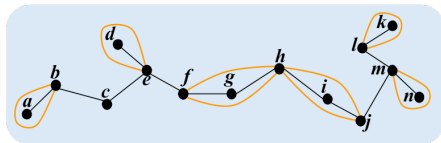
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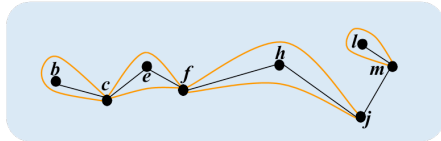
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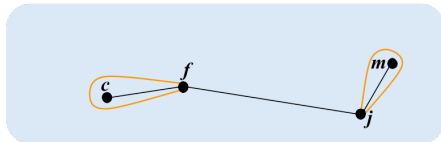
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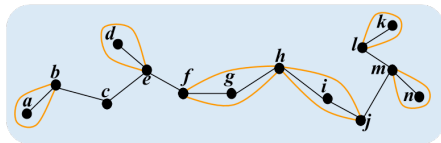
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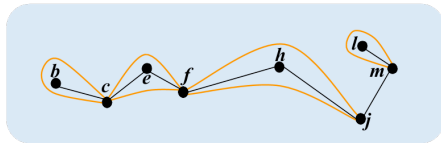
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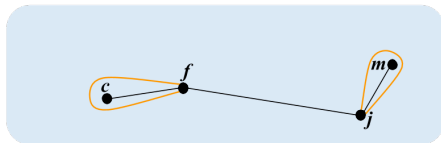
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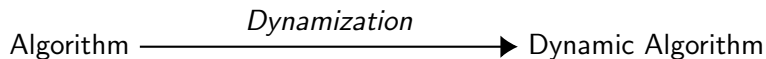


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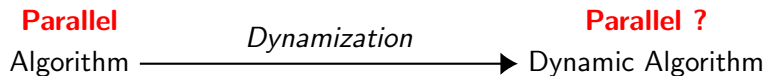
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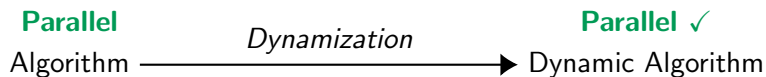
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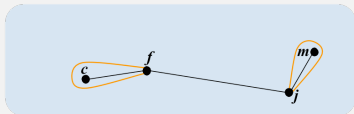
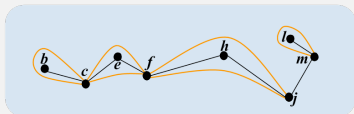
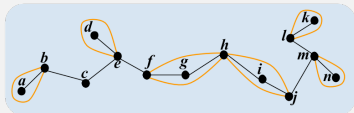


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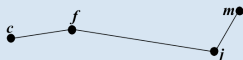
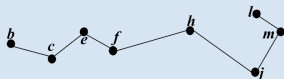
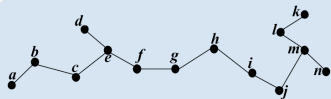
## Construction



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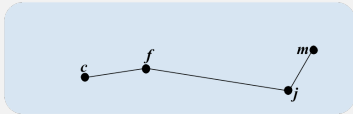
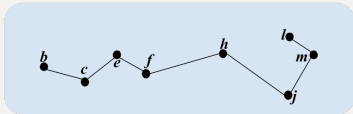
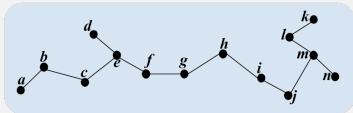
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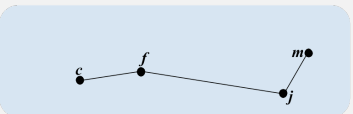
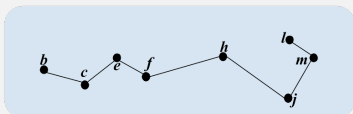
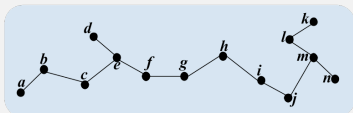
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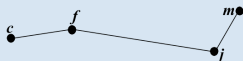
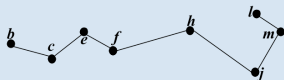
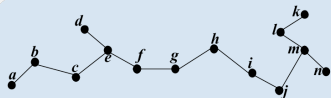
## Update



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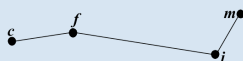
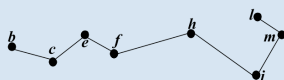
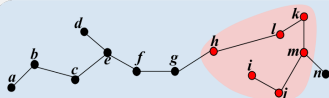
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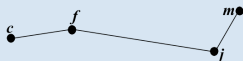
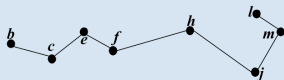
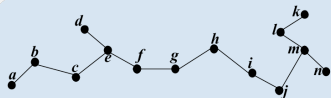
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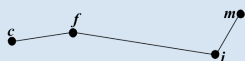
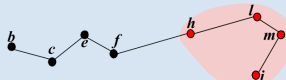
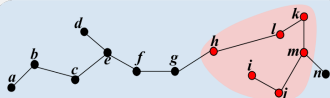
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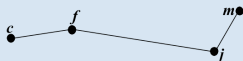
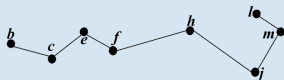
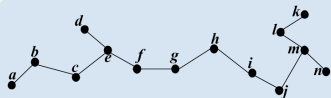


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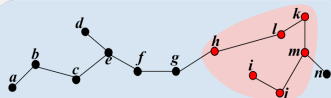
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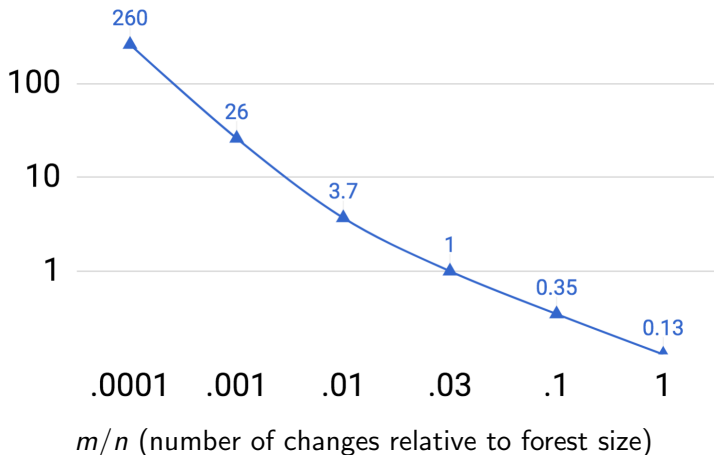


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# Measuring Performance

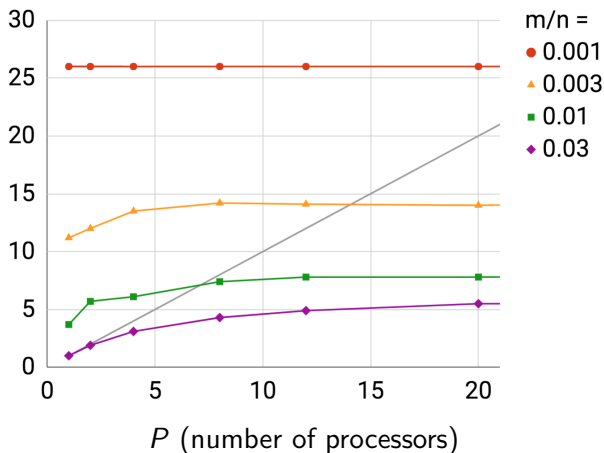
# Measuring Performance

- Work Improvement:  $\frac{T_{\text{static}}(\text{processors} = 1)}{T_{\text{update}}(\text{processors} = 1)}$
- Benefit of dynamism alone.



# Measuring Performance

- Speedup:  $\frac{T_{\text{static}}(\text{processors} = 1)}{T_{\text{update}}(\text{processors} = P)}$
- Combined benefit of dynamism and parallelism on  $P$  processors.



# Conclusion

## Summary

- We dynamized parallel tree contraction.
- The resulting algorithm is efficient both in theory and practice.

## Closing Thoughts

- Some parallel algorithms are amenable to dynamization.
  - ▶ Take advantage of independent subproblems.
  - ▶ How many more examples are there?
- Are there general purpose techniques for *automatic* dynamization?

End

Thank You!  
Questions?

# Clustering Hierarchy

