

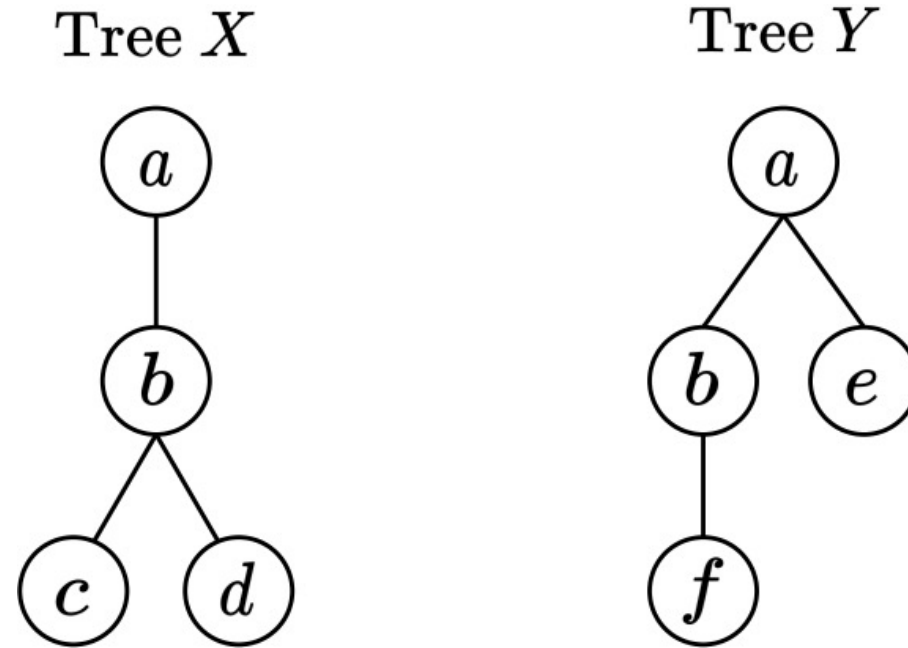
# **X-TED: Massive Parallelization of Tree Edit Distance**

**Dayi Fan<sup>†</sup>, Rubao Lee<sup>\*</sup>, Xiaodong Zhang<sup>†</sup>**

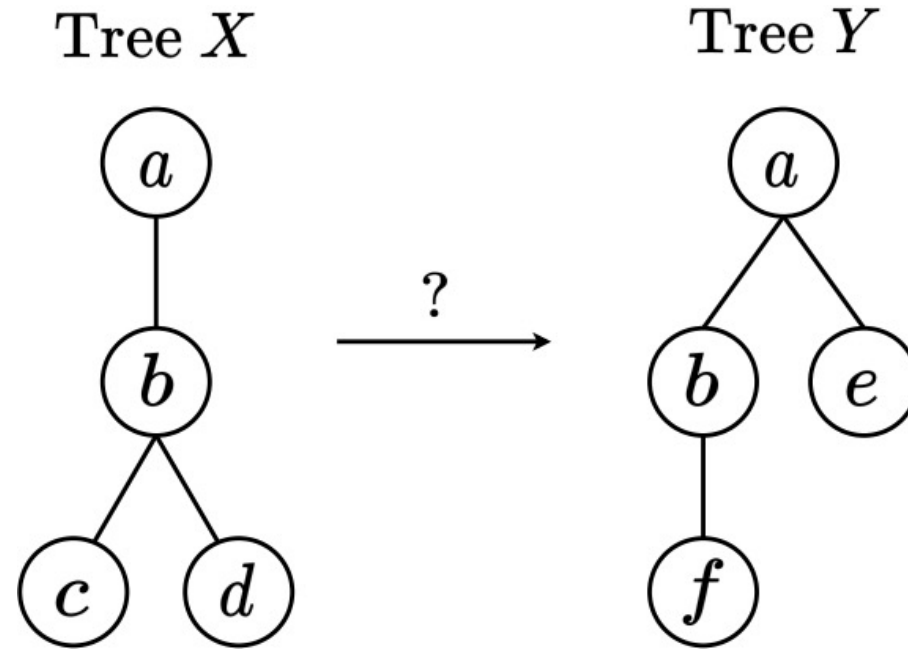
<sup>†</sup> The Ohio State University

<sup>\*</sup> Freelance

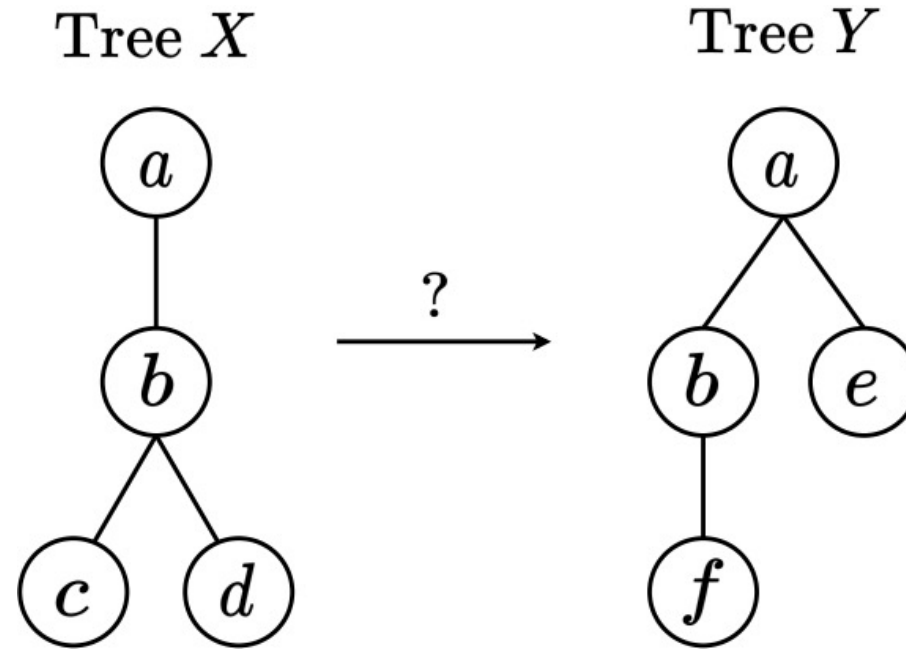
# What is Tree Edit Distance (TED)?



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# What is Tree Edit Distance (TED)?



- The **minimum cost** of transforming one tree into another by *edit operations*

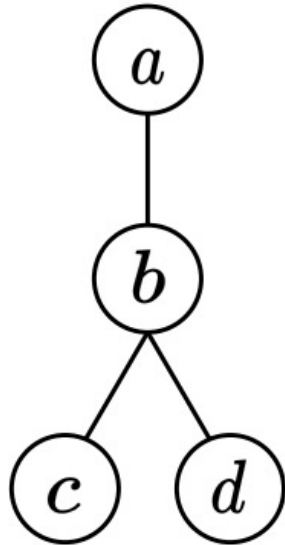
# Tree Edit Distance (TED)

- Three edit operation types: delete, insert, rename

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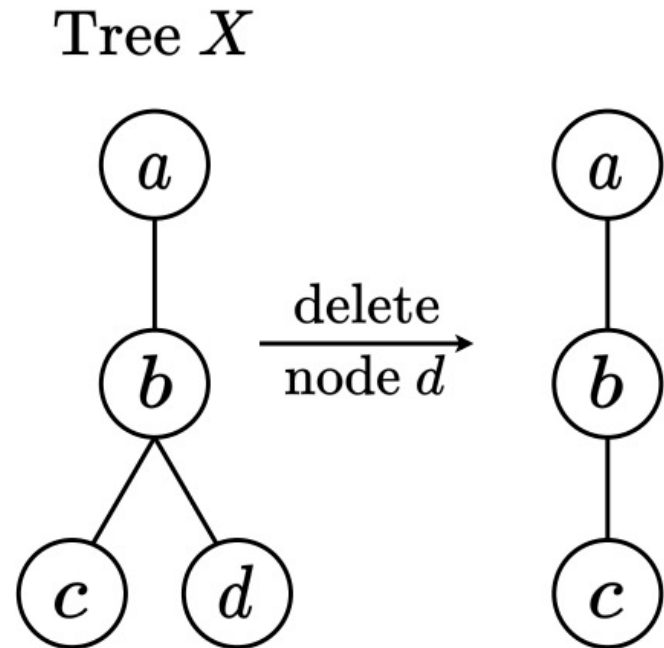
- Three edit operation types: delete, insert, rename

Tree  $X$



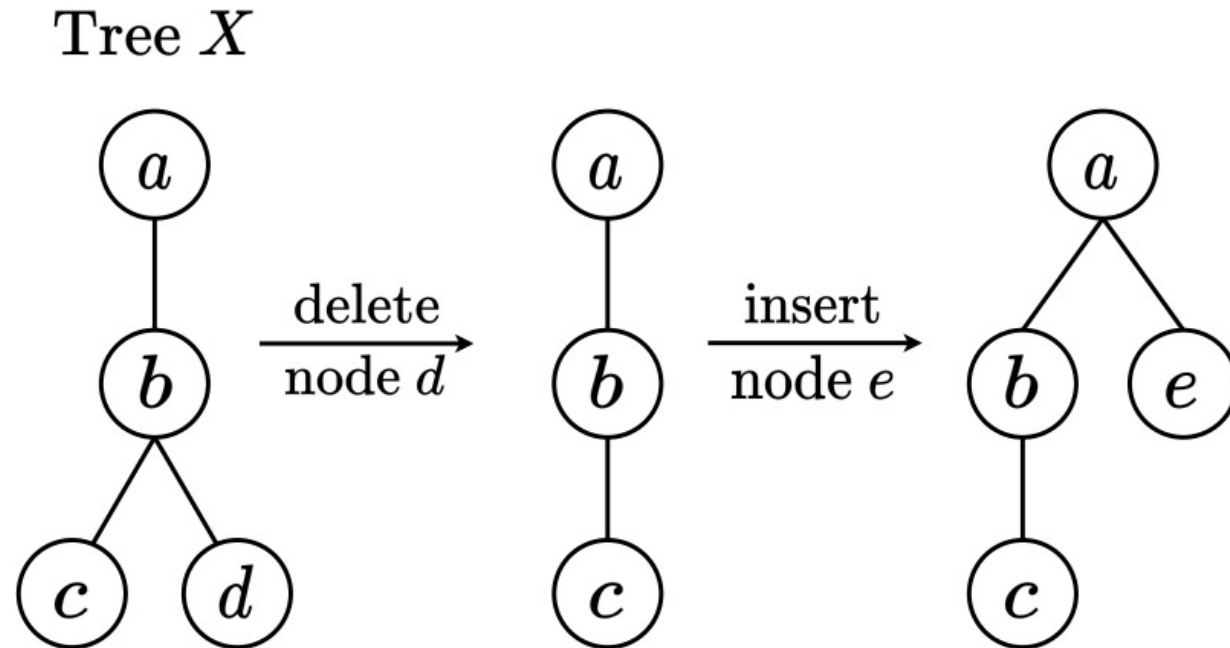
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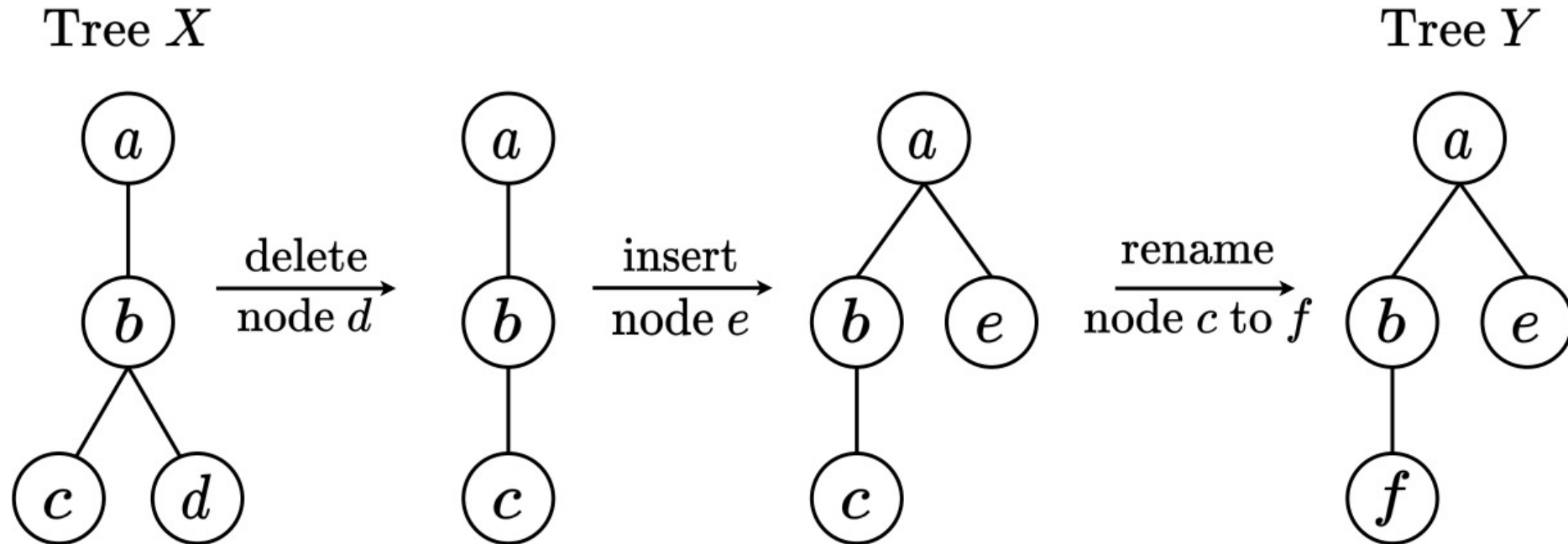
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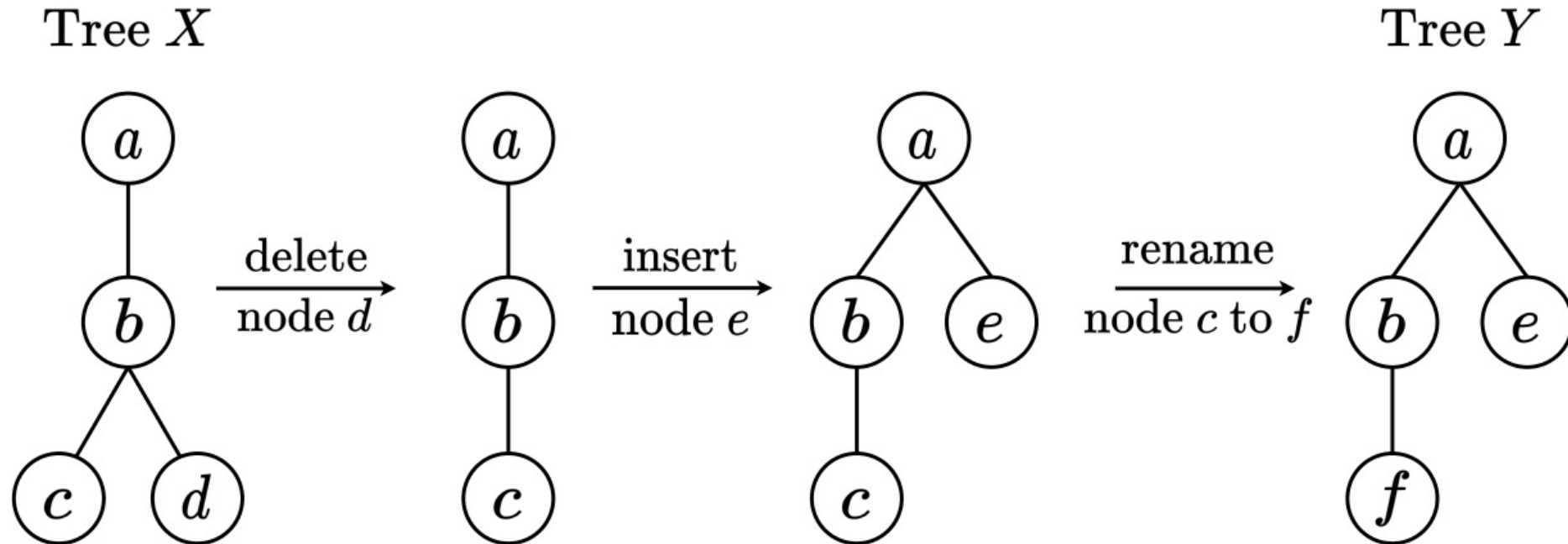
# Tree Edit Distance (TED)

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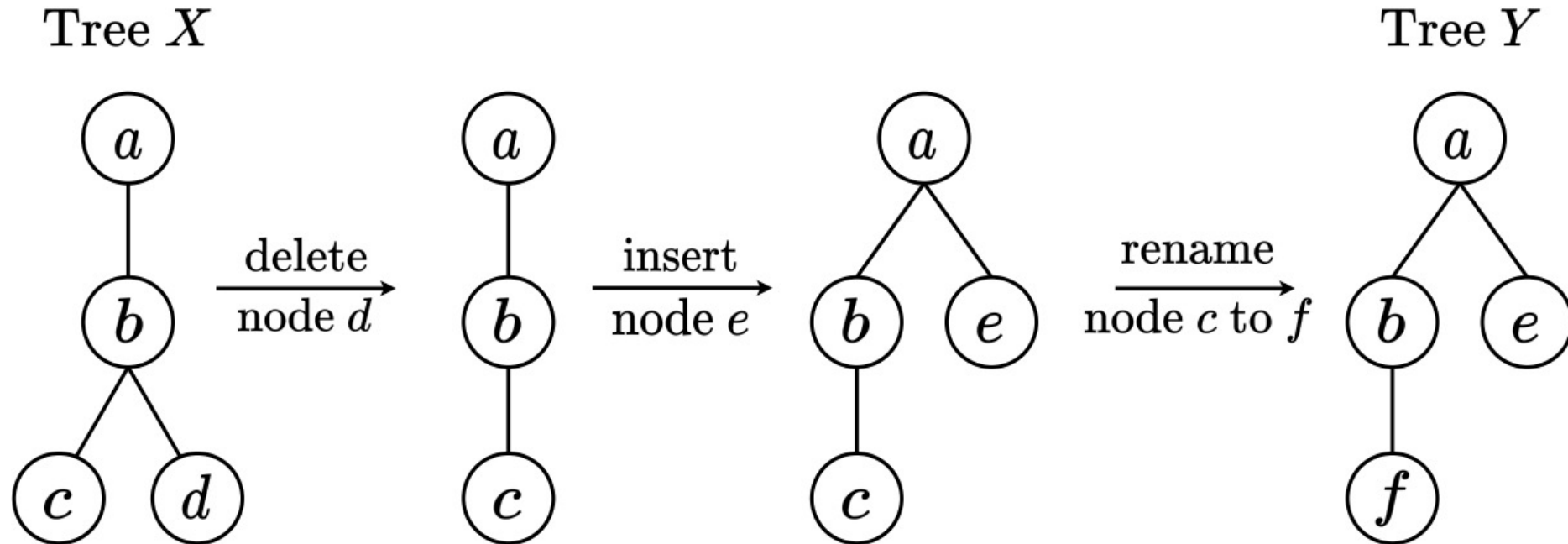
# Tree Edit Distance (TED)

- If the cost of each edit operation is 1, ...



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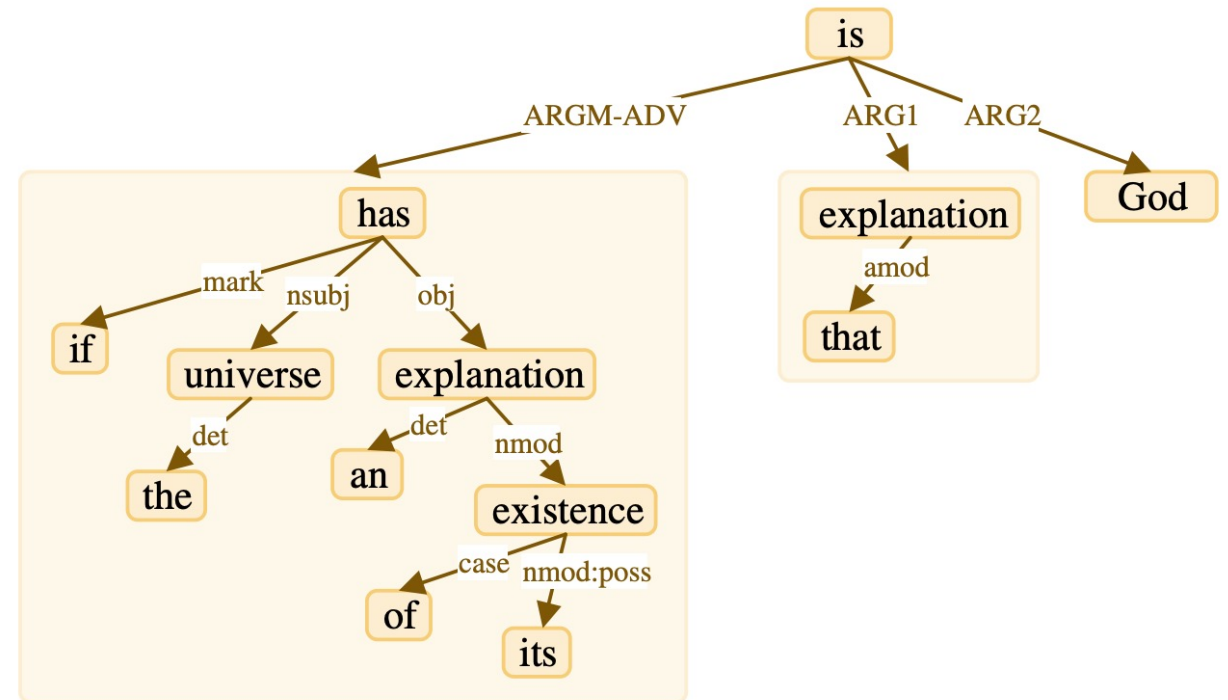


$$\text{TED}(X, Y) = 3$$

# TED has a wide range of applications

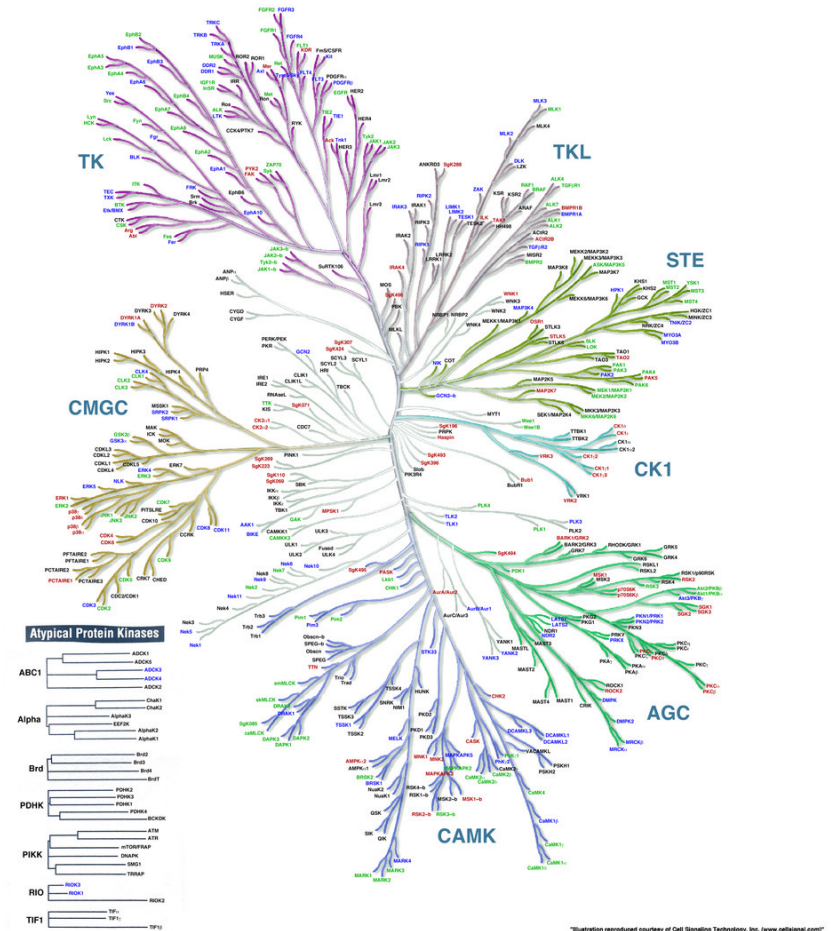
# TED has a wide range of applications

- Natural language processing
  - AI Generated Content



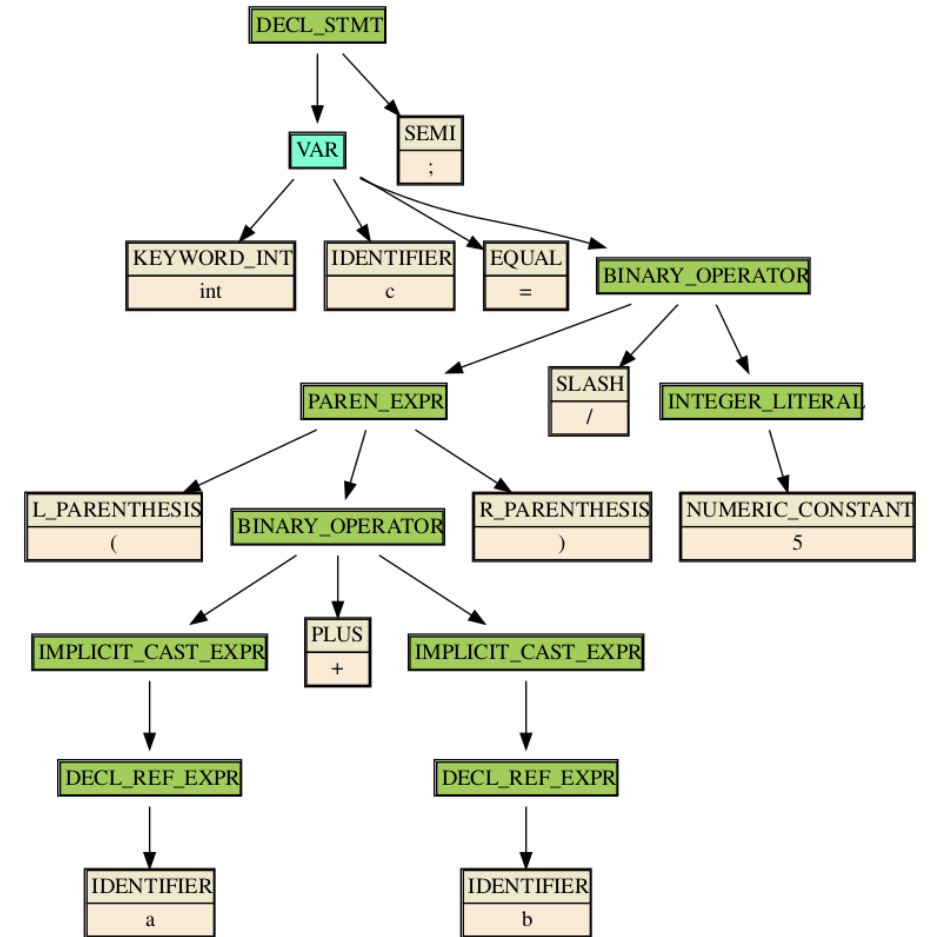
# TED has a wide range of applications

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  - Protein structure and evolutionary



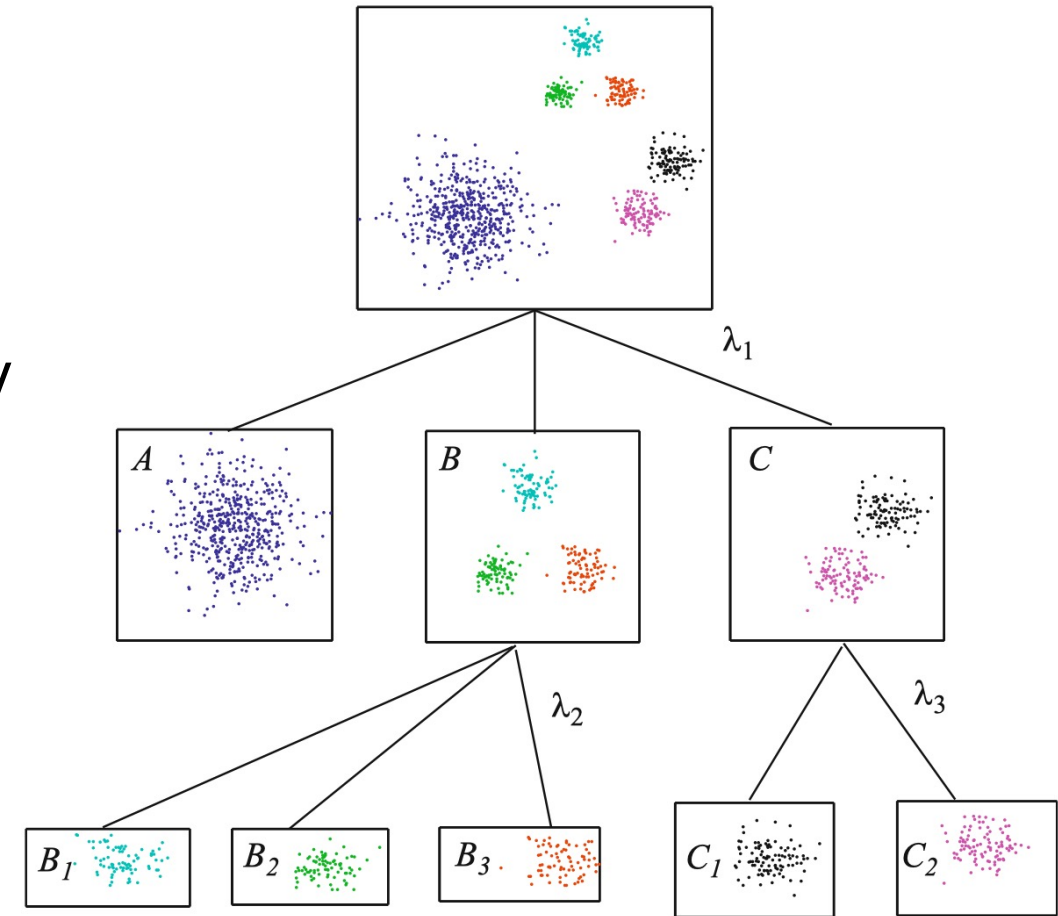
# TED has a wide range of applications

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- Bioinformatics
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- Software Engineering
  - Code similarity detection



# TED has a wide range of applications

- Natural language processing
  - AI Generated Content
- Bioinformatics
  - Protein structure and evolutionary
- Software Engineering
  - Code similarity detection
- Machine Learning
  - Classifications and clustering





- The first TED algorithm (1979) with complexity  $O(n^6)$

# The Basic TED Algorithm

- The most widely-used TED algorithm (1989)

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- Dynamic programming (DP)

# The Basic TED Algorithm

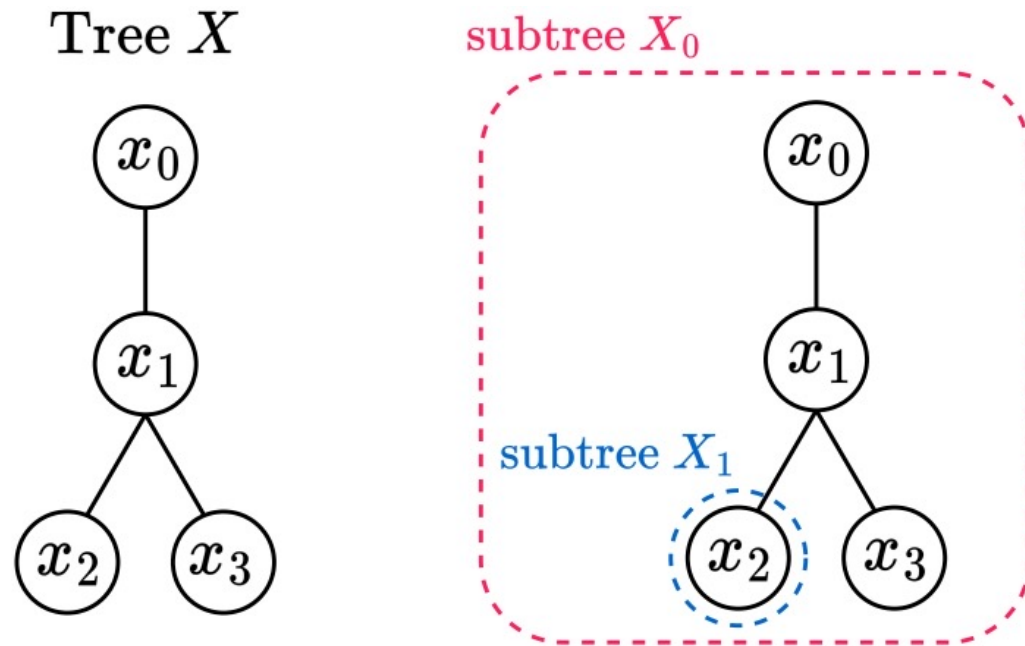
- The most widely-used TED algorithm (1989)
- Dynamic programming (DP)
- The worst-case complexity:  $O(n^4)$

# The Basic TED Algorithm

- Step 1 – divide tree to subtrees

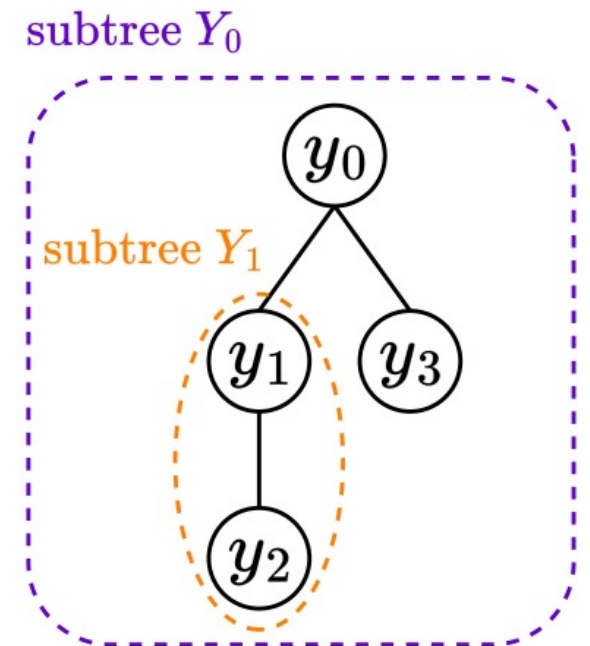
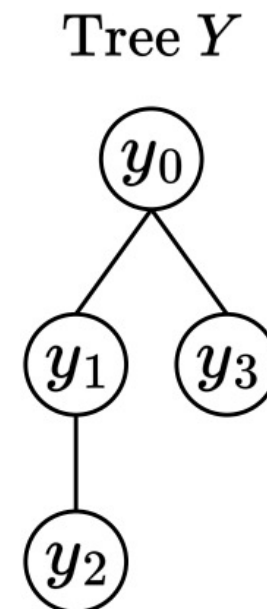
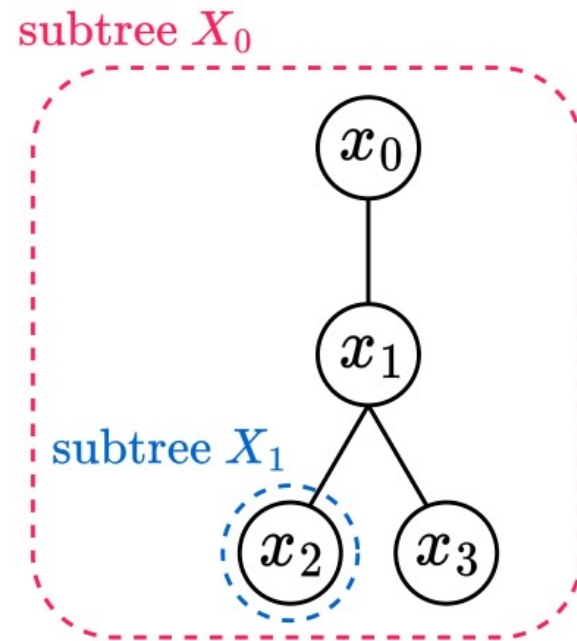
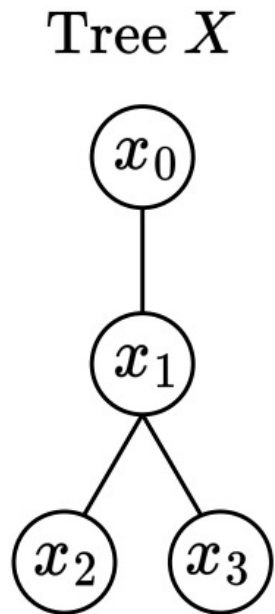
# The Basic TED Algorithm

- Step 1 – divide tree to subtrees



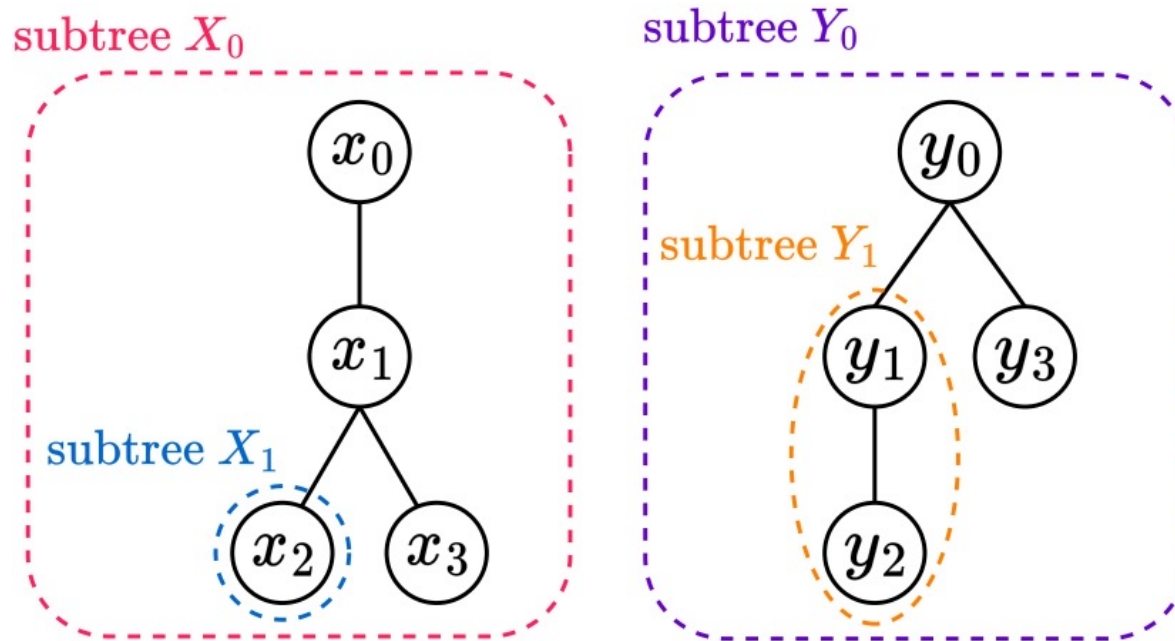
# The Basic TED Algorithm

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# The Basic TED Algorithm

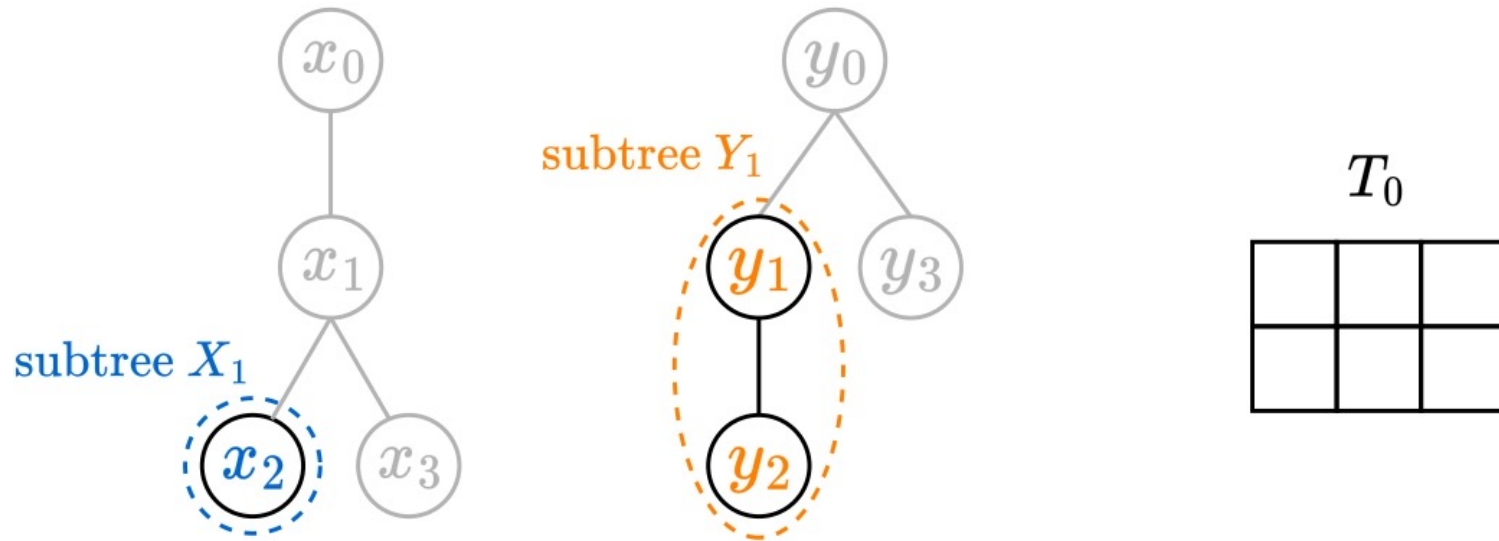
- Step 2 – DP tables to compute distance for each subtree pair





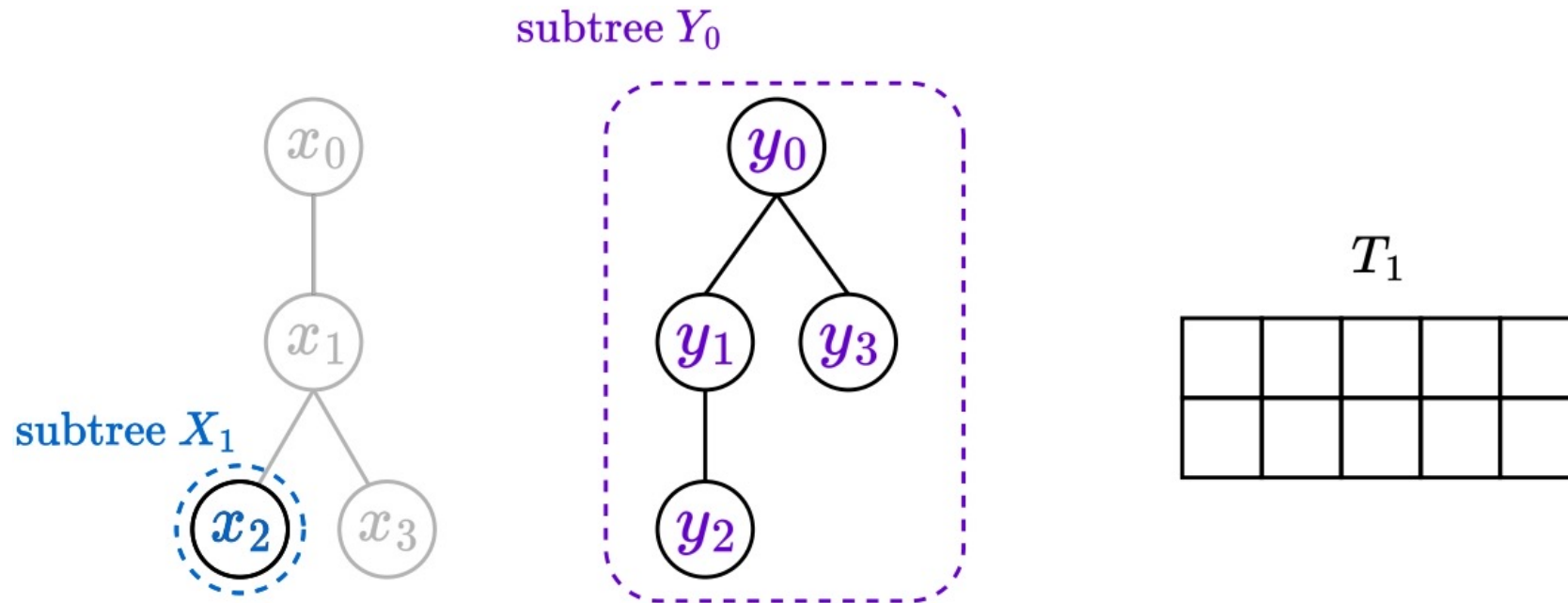
# The Basic TED Algorithm

- 1<sup>st</sup> Table: DP table for subtree  $X_1$  and subtree  $Y_1$



# The Basic TED Algorithm

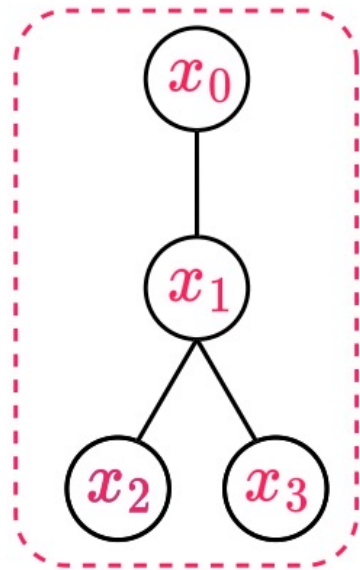
- 2<sup>nd</sup> Table: DP table for subtree  $X_1$  and subtree  $Y_0$



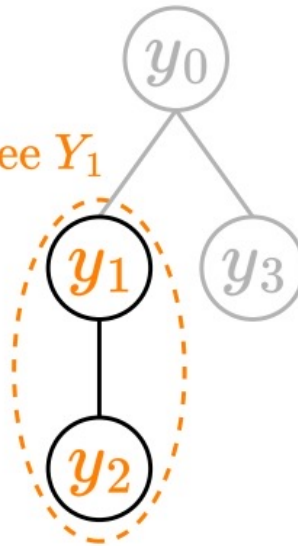
# The Basic TED Algorithm

- 3<sup>rd</sup> Table: DP table for subtree  $X_0$  and subtree  $Y_1$

subtree  $X_0$



subtree  $Y_1$

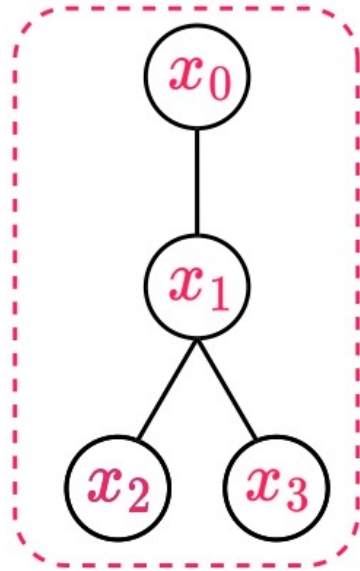


$T_2$

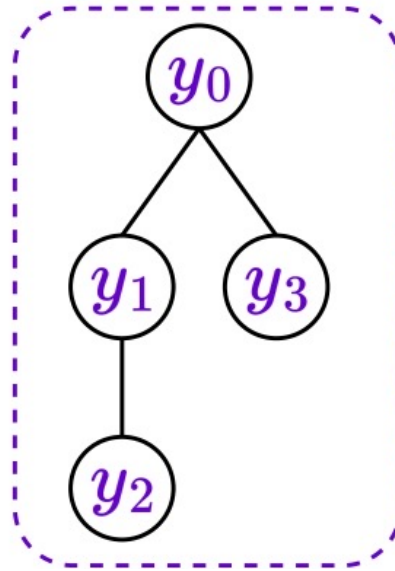

# The Basic TED Algorithm

- 4<sup>th</sup> Table: DP table for subtree  $X_0$  and subtree  $Y_0$

subtree  $X_0$



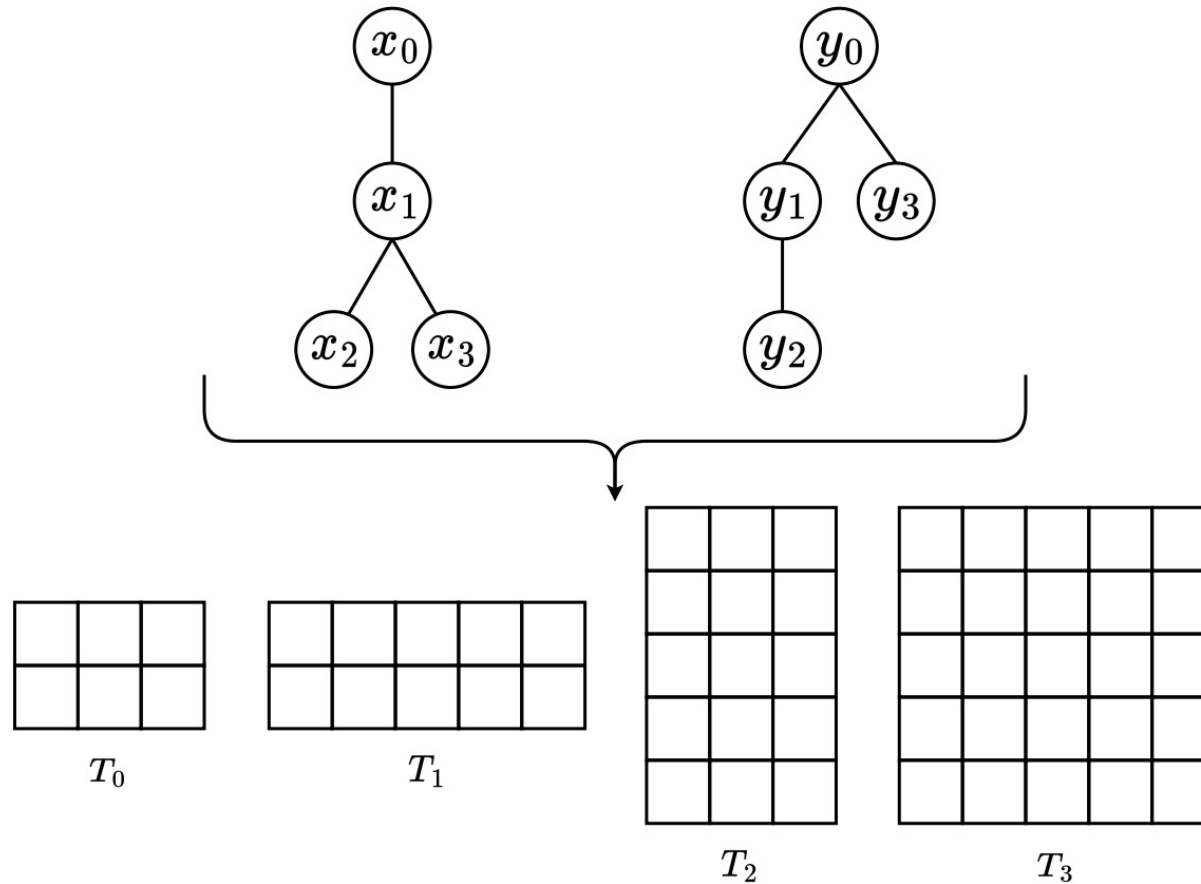
subtree  $Y_0$



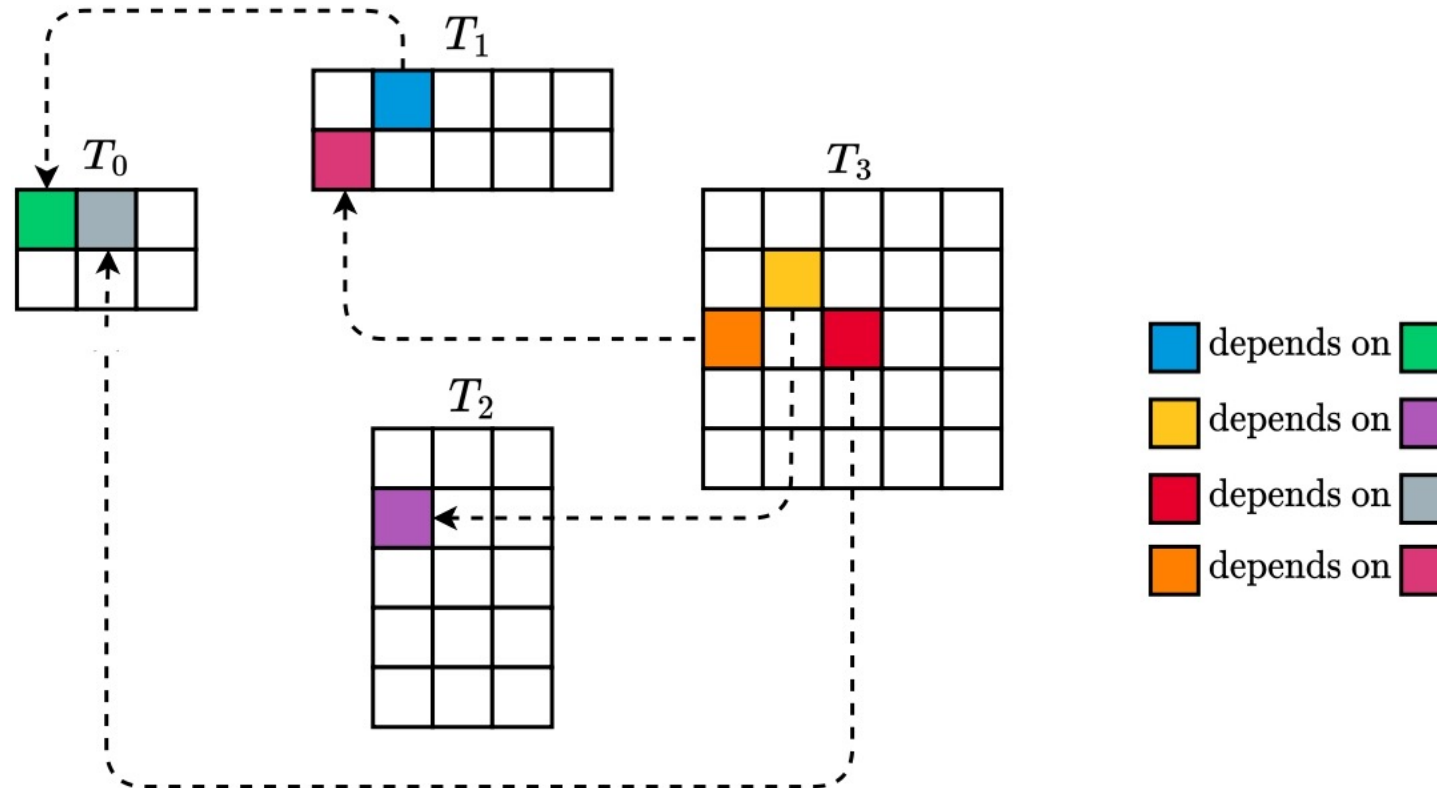
$T_3$


# The Basic TED Algorithm

- Step 3 – return TED result after all DP tables are computed



# Data Dependencies Example



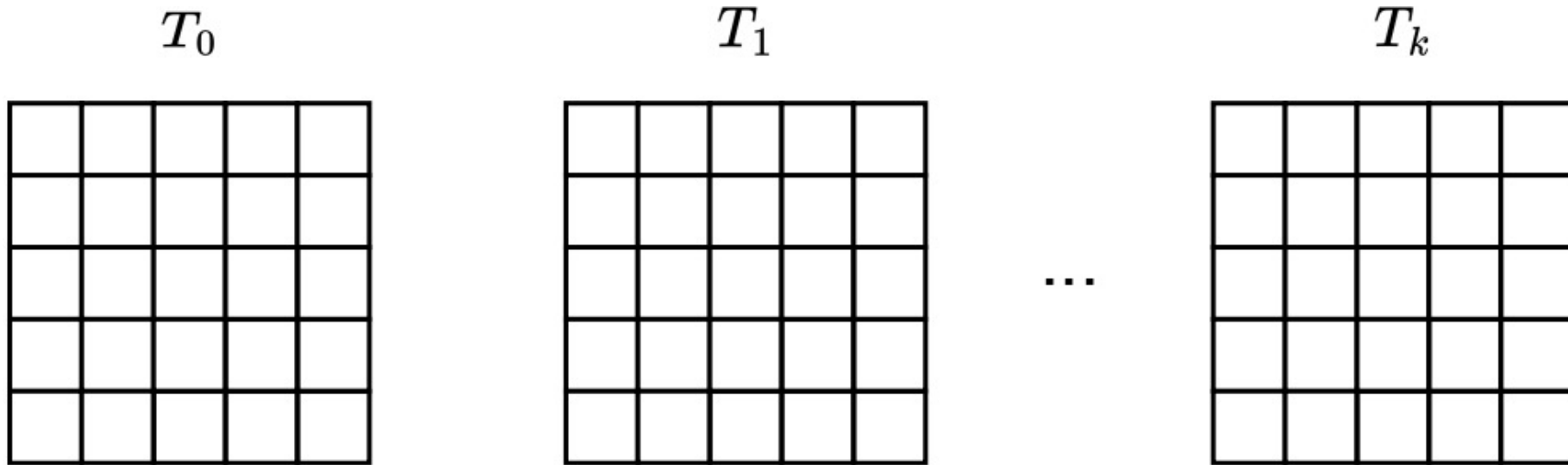
- Hinder parallel processing

# Existing Parallel Solution

- Wave-front parallel computing

# Existing Parallel Solution

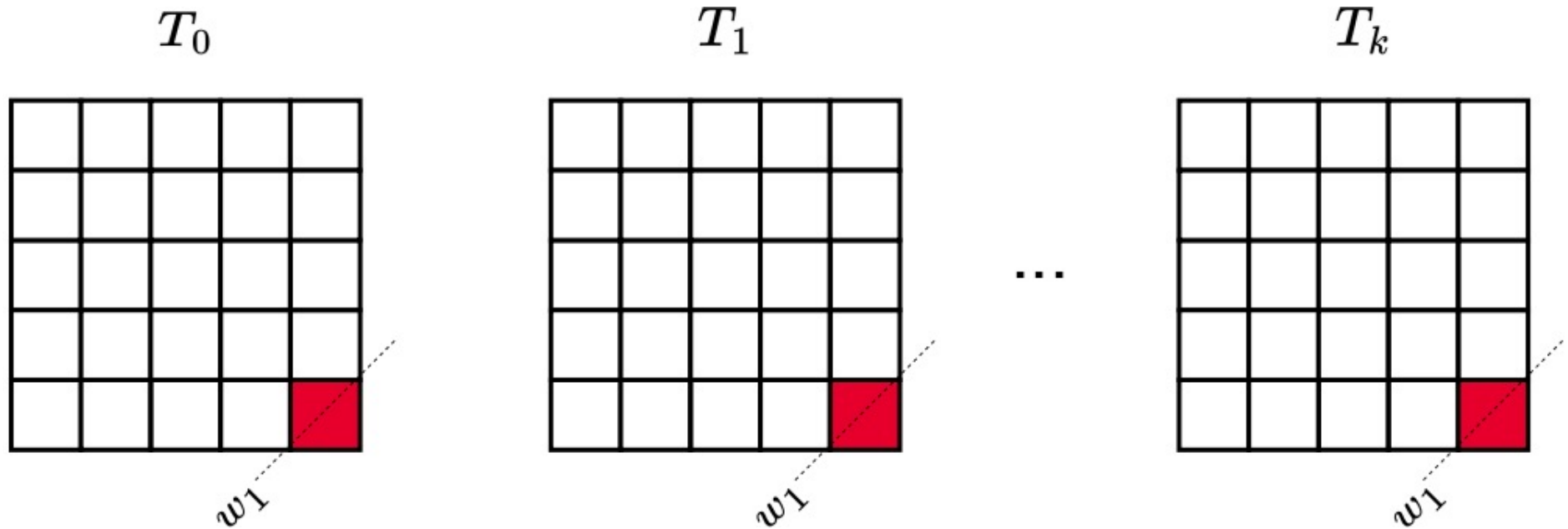
- Wave-front parallel computing





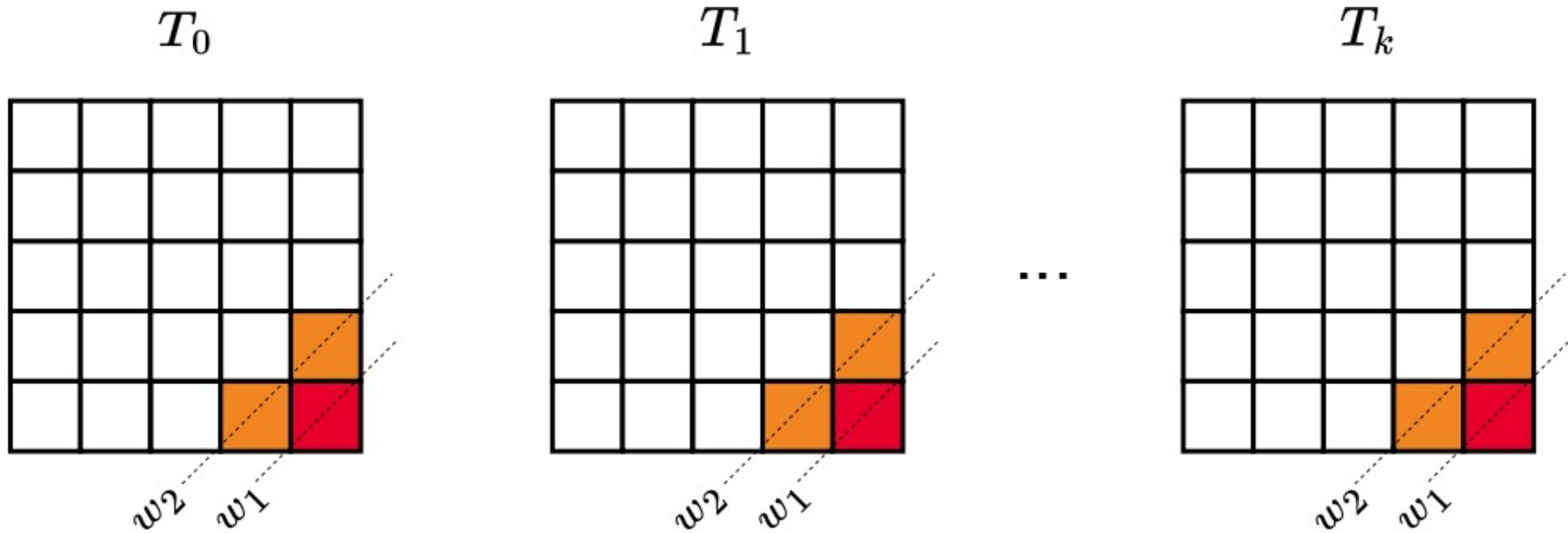
# Existing Parallel Solution

- First, compute red units for all tables in parallel



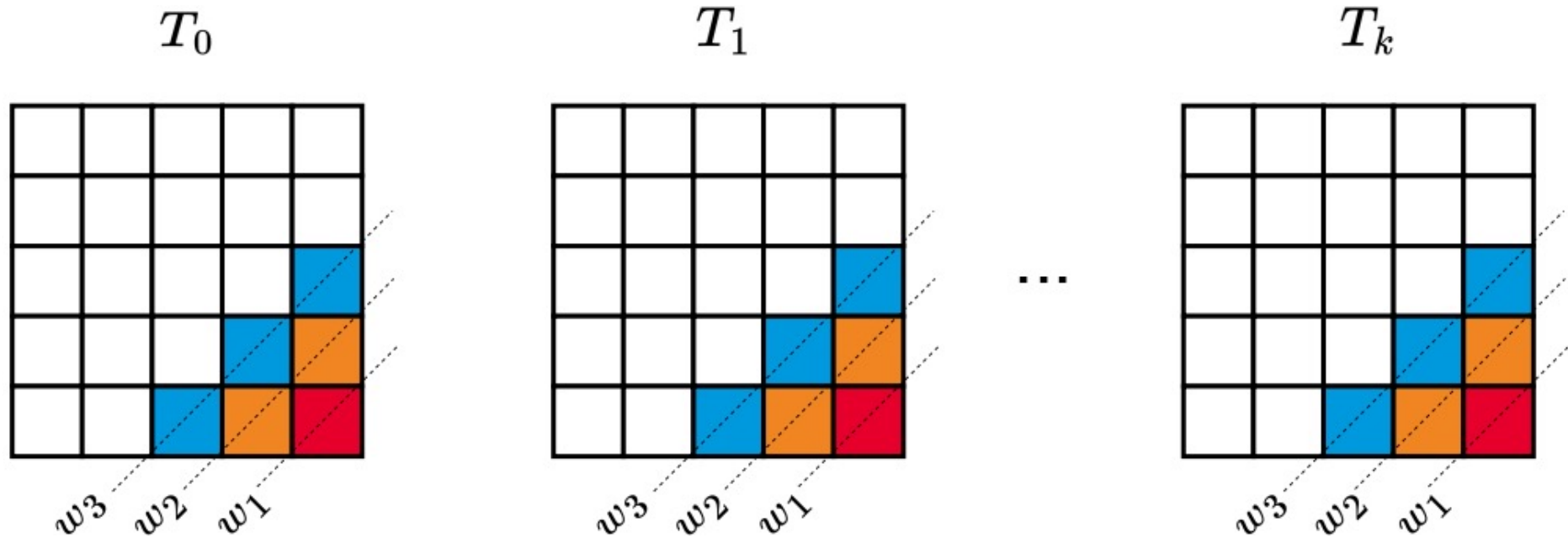
# Existing Parallel Solution

- Then, compute orange units for all tables in parallel



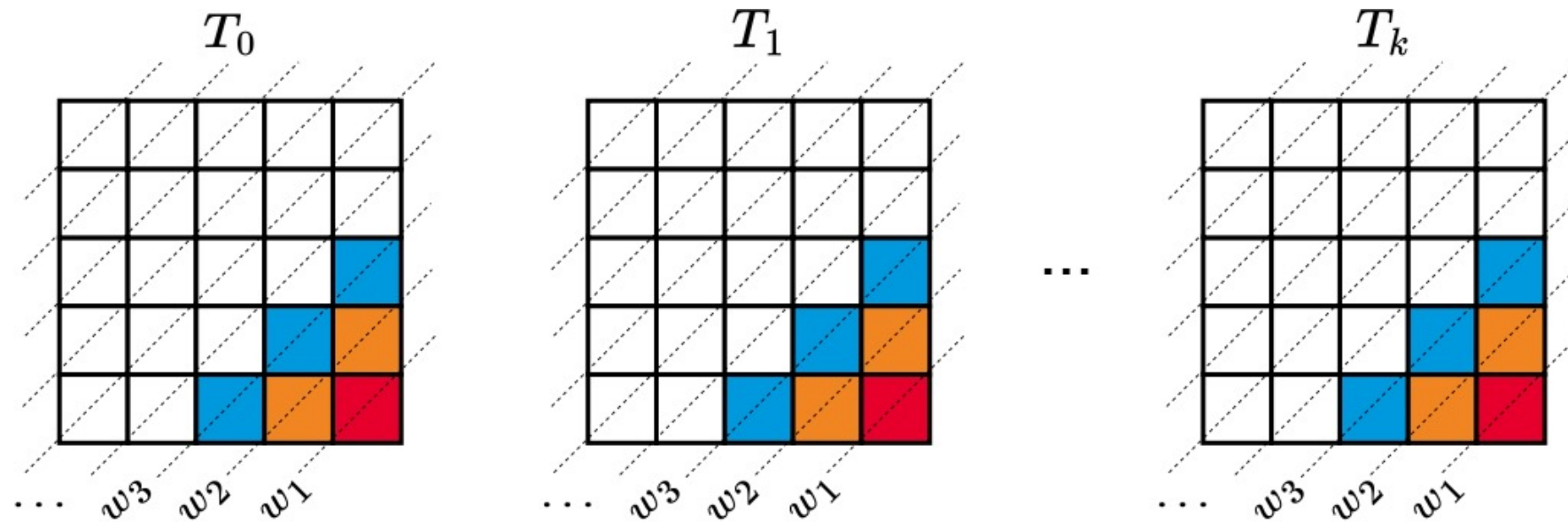
# Existing Parallel Solution

- Next, compute blue units for all tables in parallel



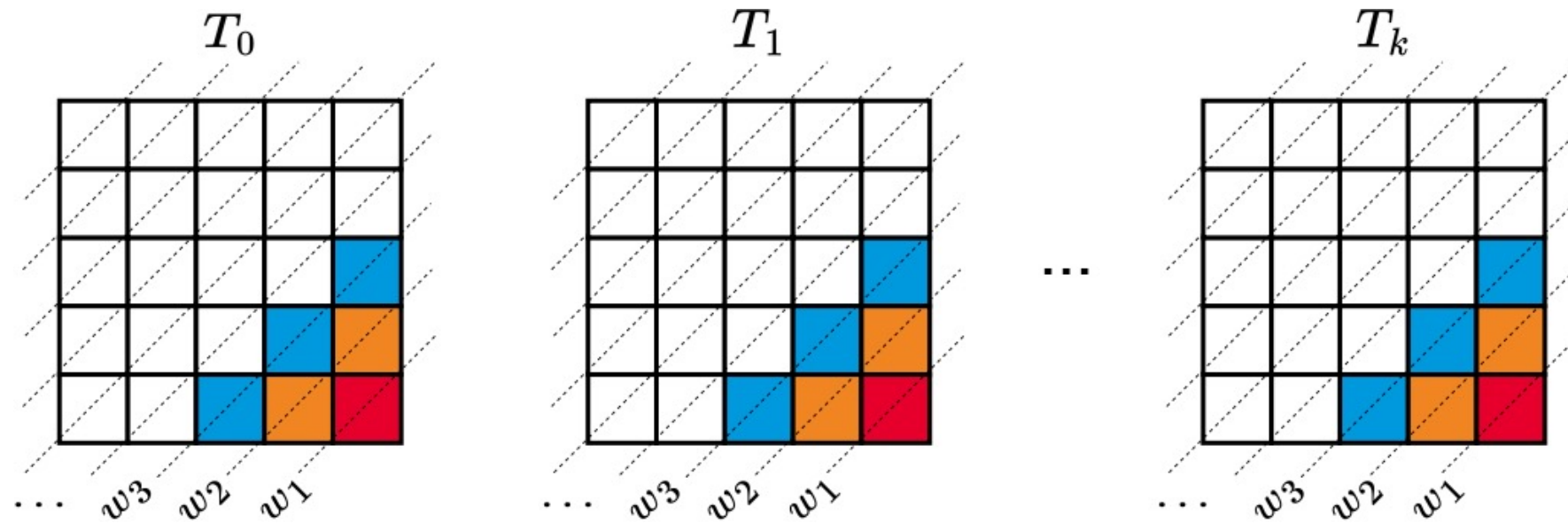


# Existing Parallel Solution



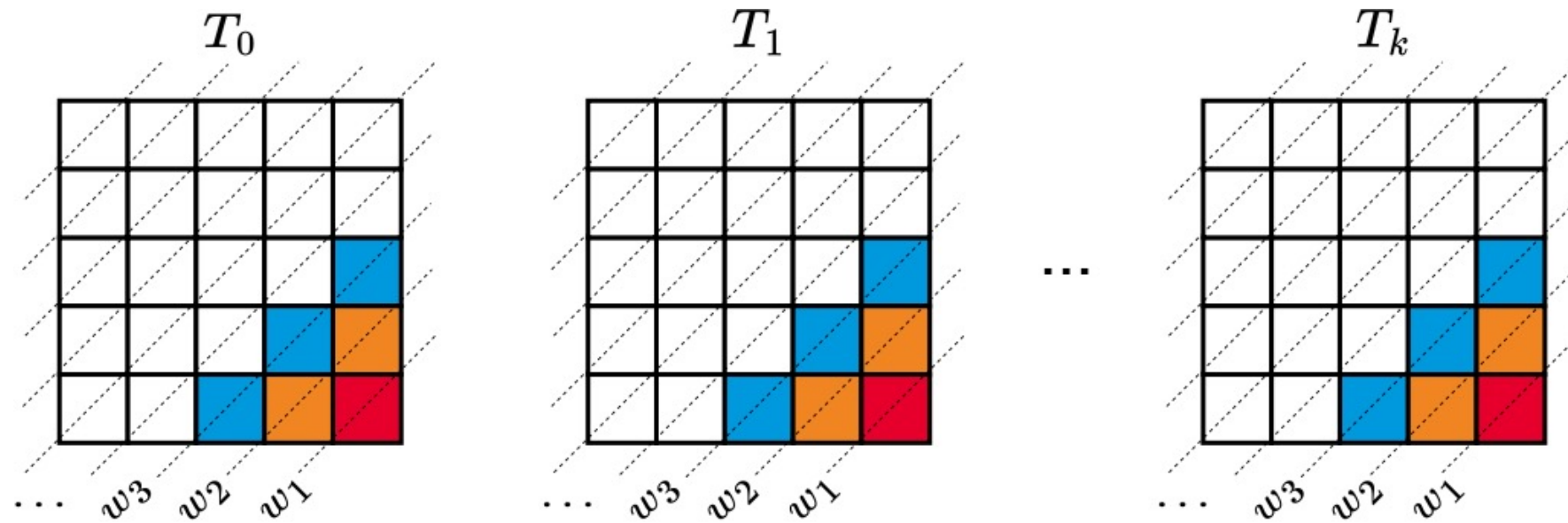
- A huge memory space

# Existing Parallel Solution



- A huge memory space
- Frequent synchronizations

# Existing Parallel Solution



- A huge memory space
- Frequent synchronizations
- Load-imbalanced

# Insights from DP Patterns



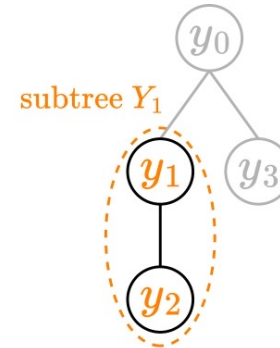
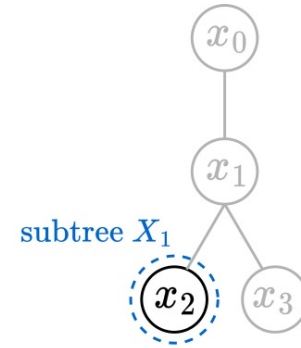
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- The computing of one DP table is viewed as a task

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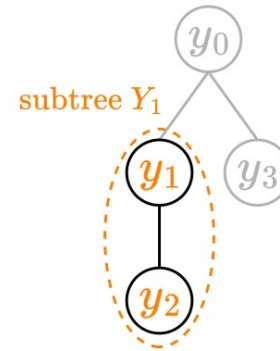
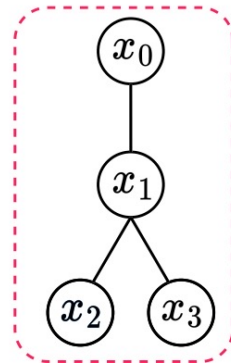
- The computing of one DP table is viewed as a task
- Data dependencies among tables are ***determined by tree structure***

# Insights from DP Patterns



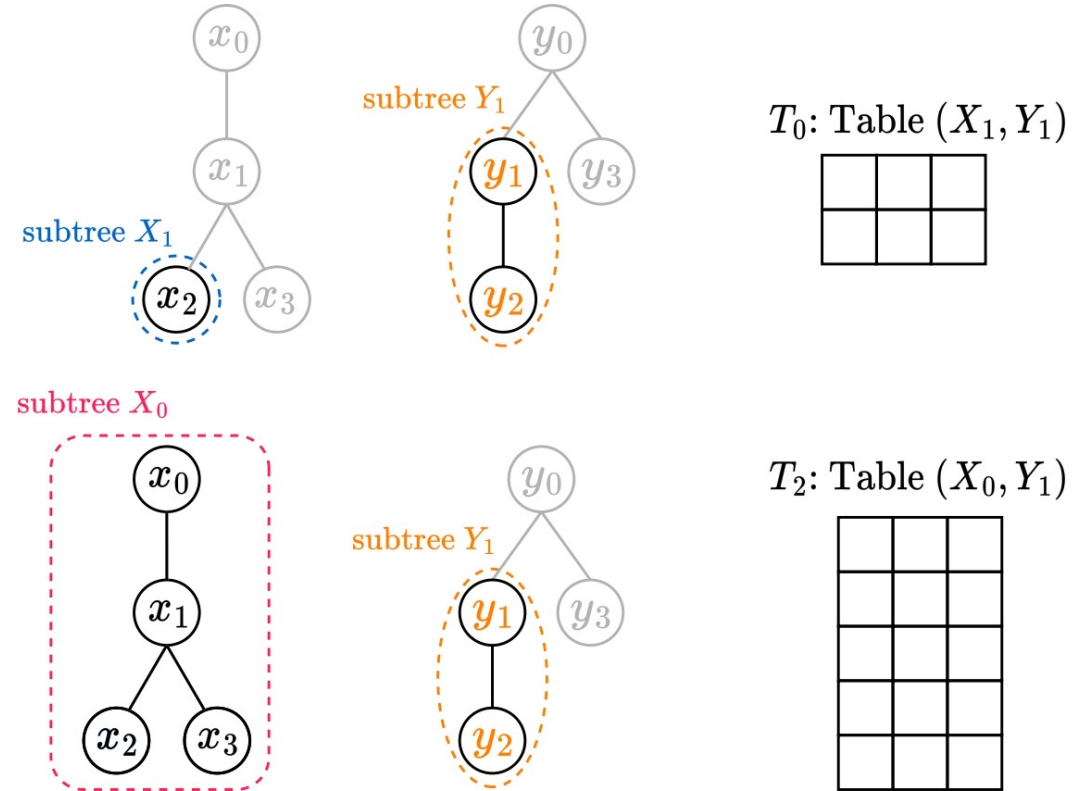
$T_0$ : Table ( $X_1, Y_1$ )


subtree  $X_0$



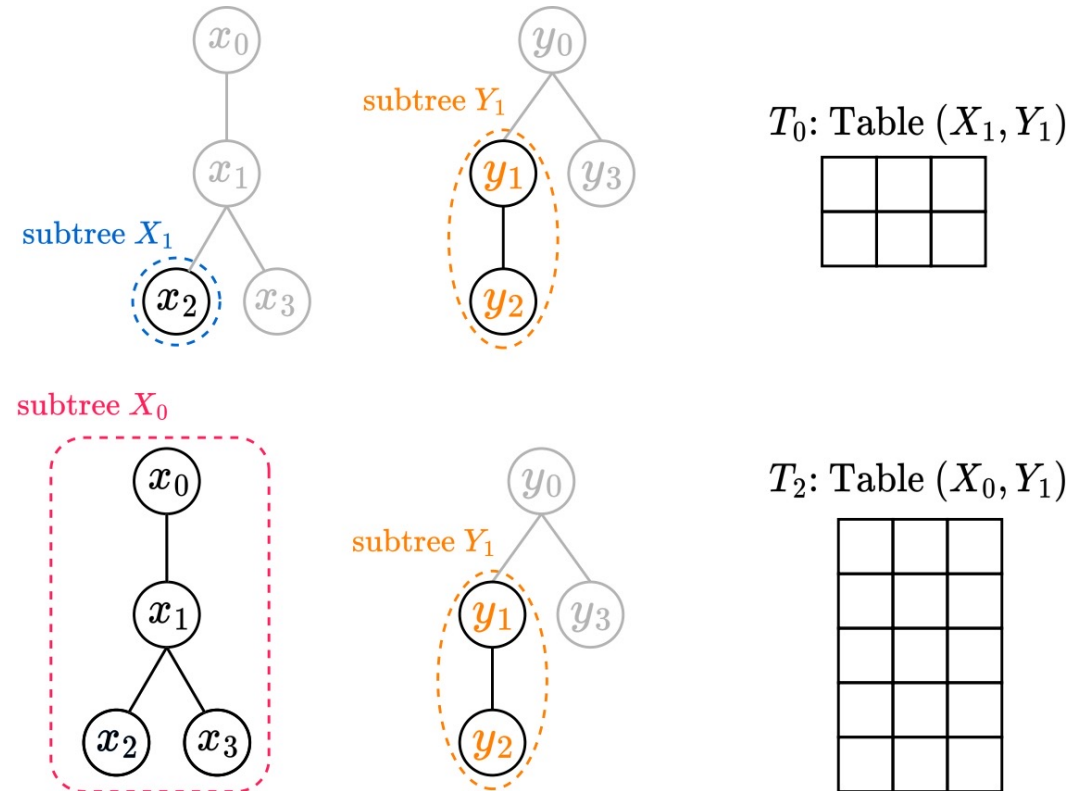
$T_2$ : Table ( $X_0, Y_1$ )


# Insights from DP Patterns



- subtree  $X_0$  contains  $X_1$

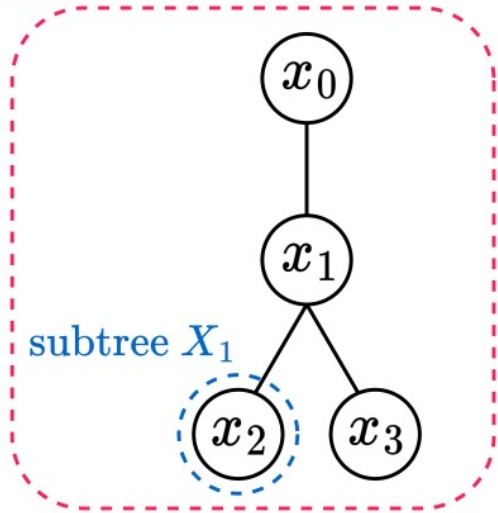
# Insights from DP Patterns



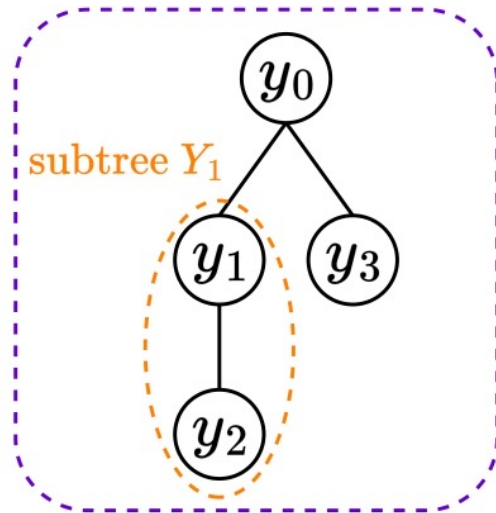
- subtree  $X_0$  contains  $X_1$   $\longrightarrow$  Table  $(X_0, Y_1)$  depends on Table  $(X_1, Y_1)$

# Insights from DP Patterns

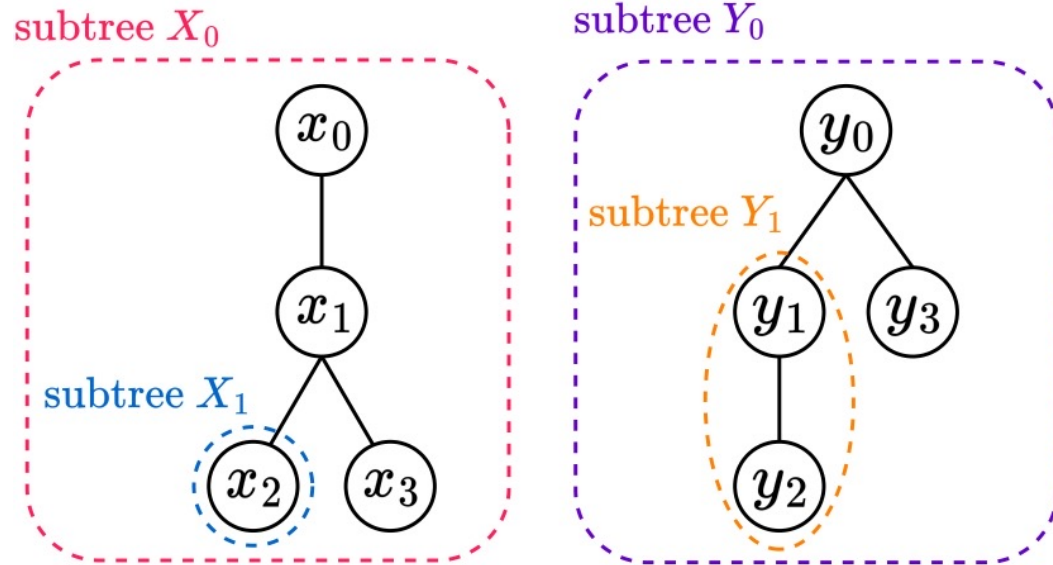
subtree  $X_0$



subtree  $Y_0$

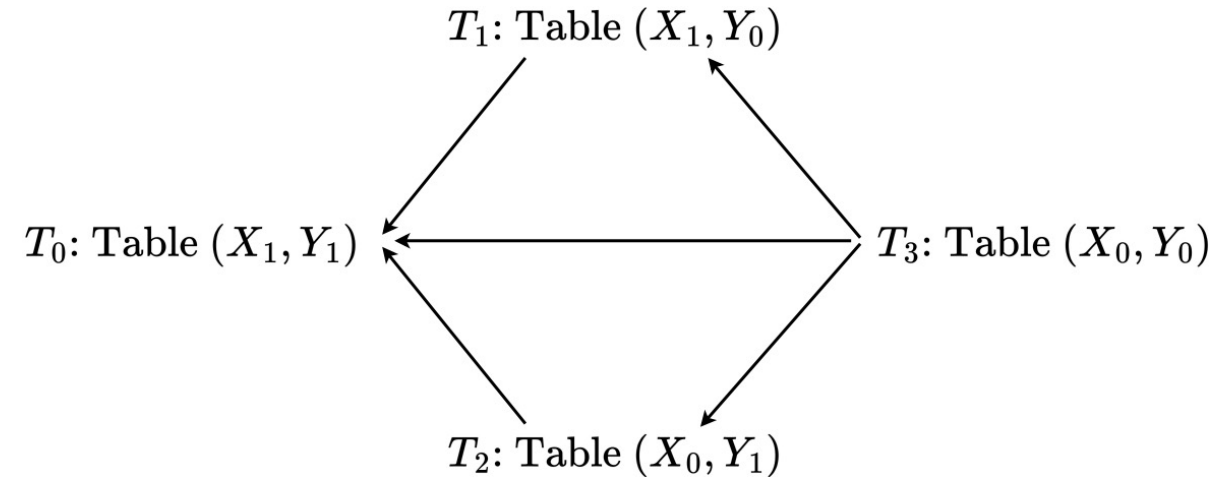
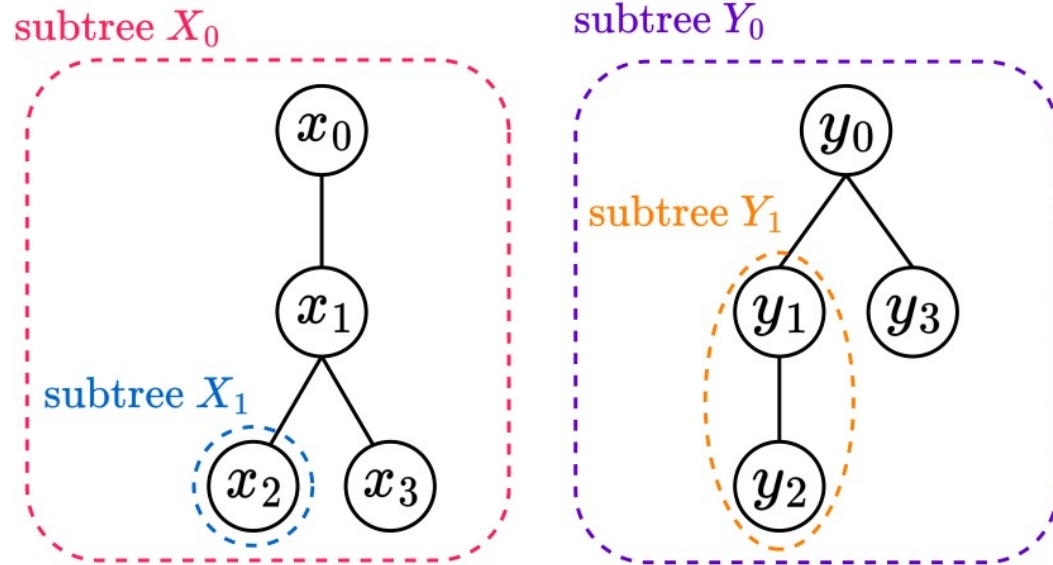


# Insights from DP Patterns



- subtree  $X_0$  contains  $X_1$ , subtree  $Y_0$  contains  $Y_1$

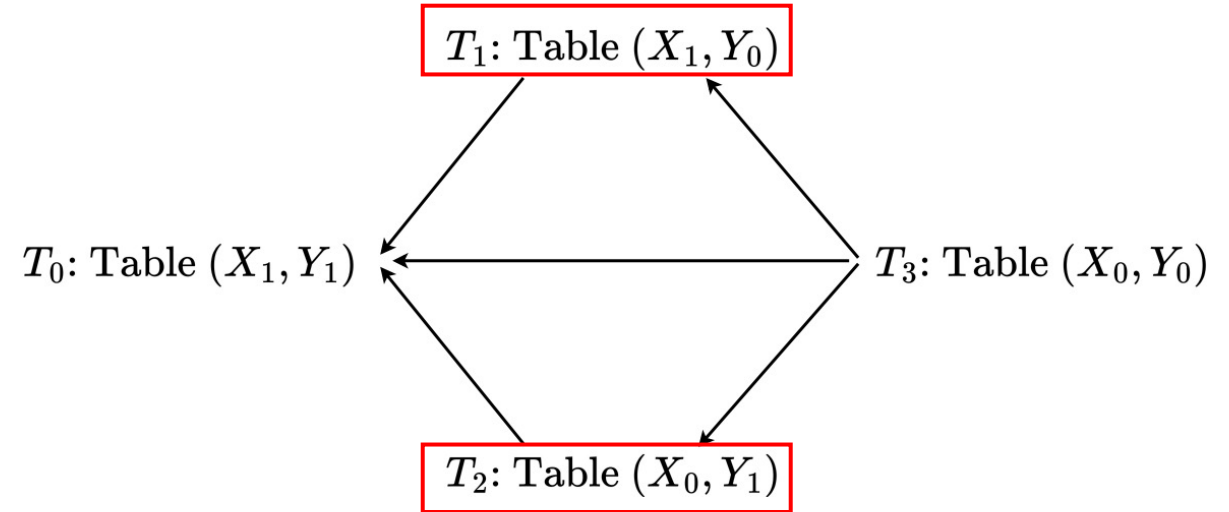
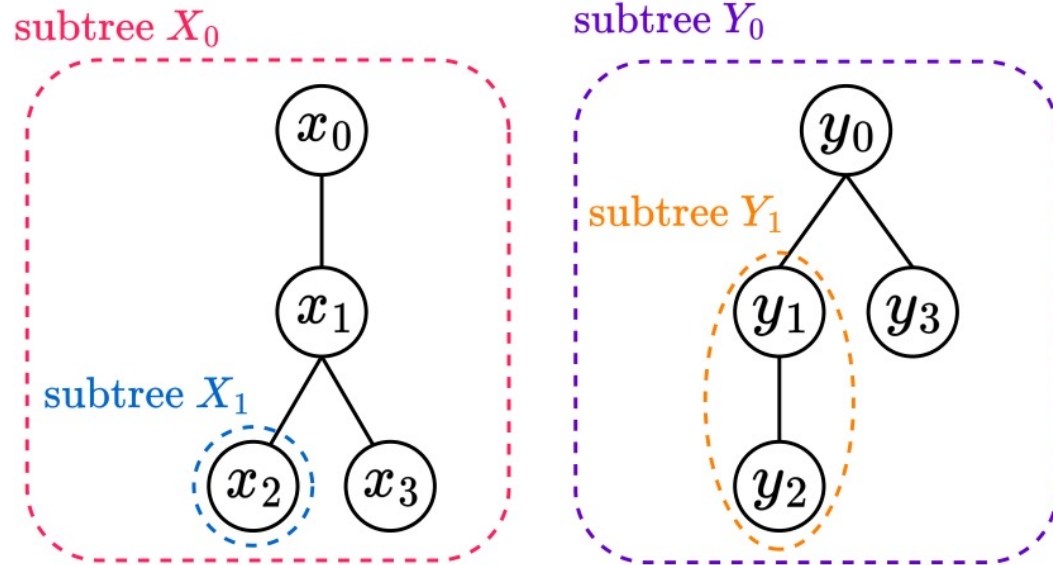
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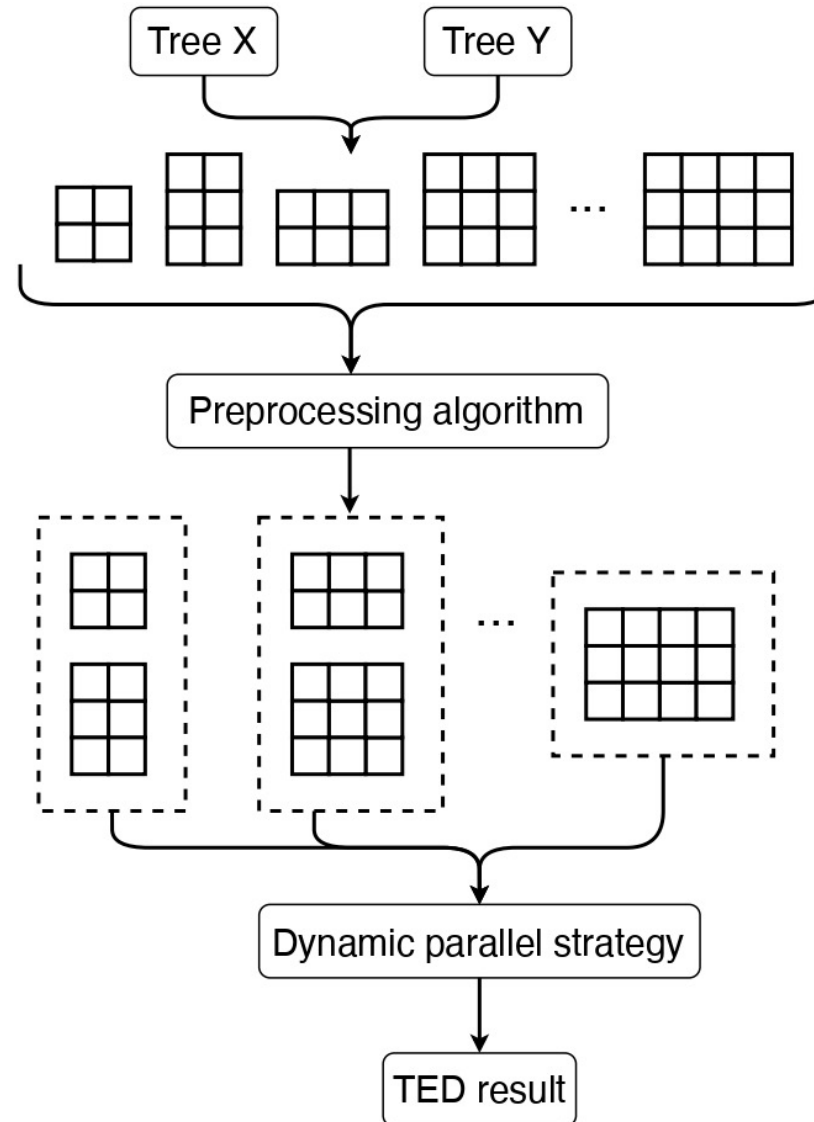
# Insights from DP Patterns



- $T_1$  and  $T_2$  then can be computed in parallel

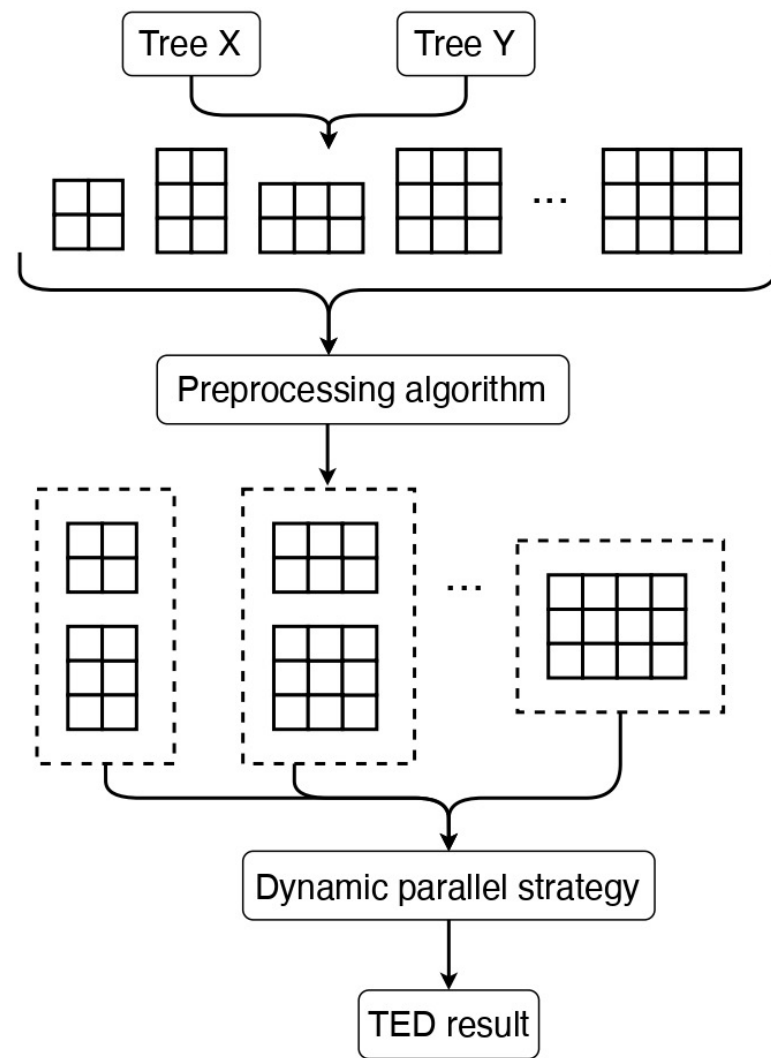
# Our Solution: **X-TED** for Massively Parallel Computing

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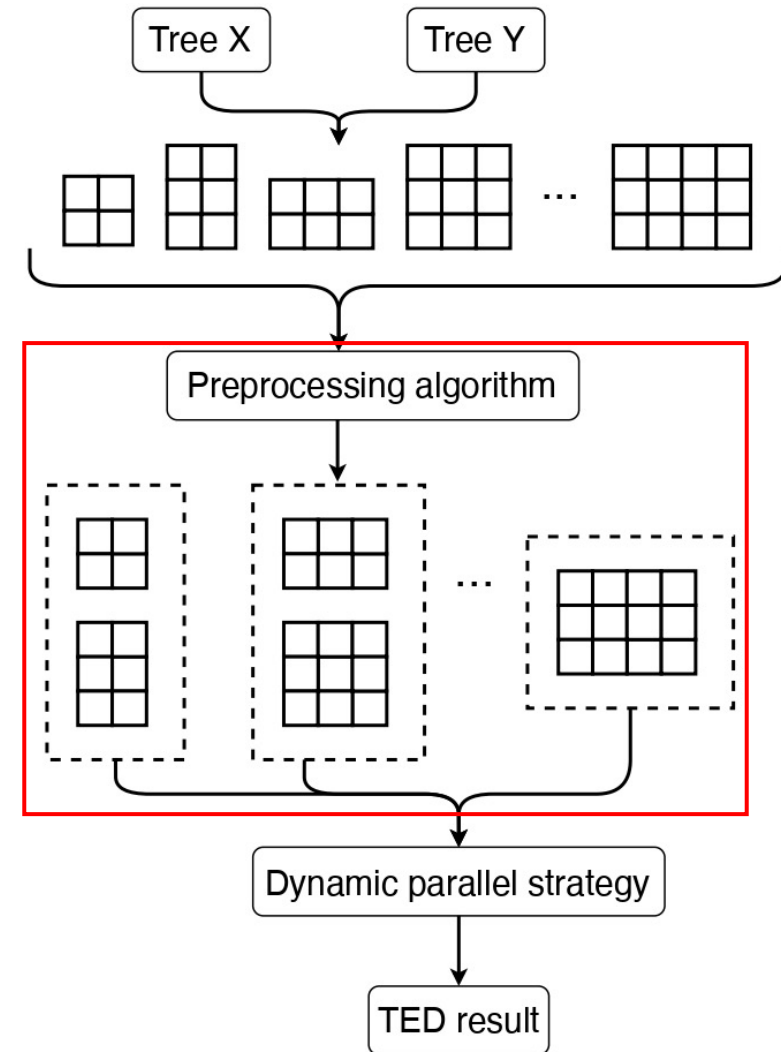
# X-TED

- Memory saving



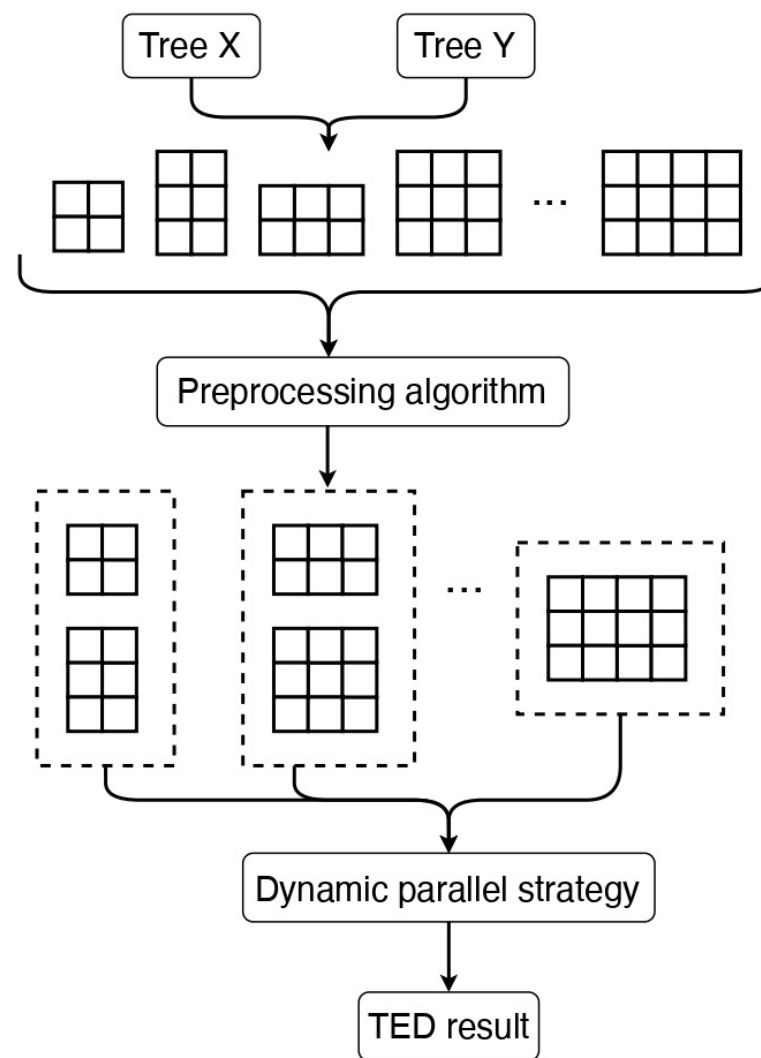
# X-TED

- Memory saving
  - Each table is a task
  - Each processor only stores one table



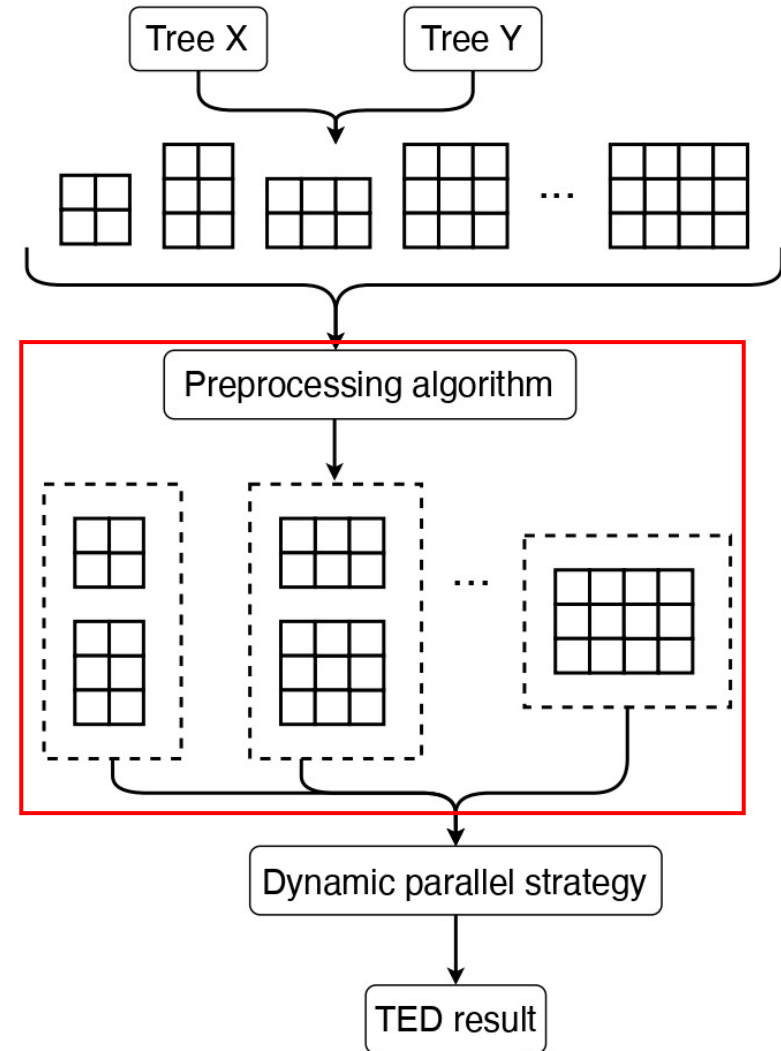
# X-TED

- Less synchronizations



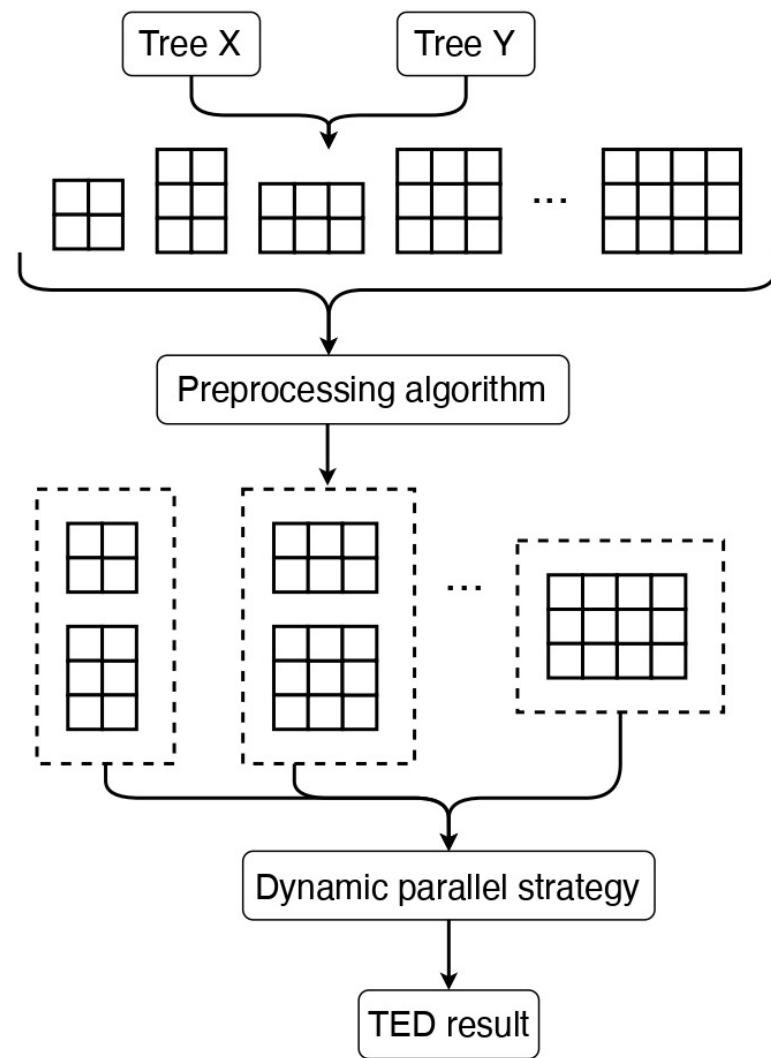
# X-TED

- Less synchronizations
  - Only sync. once at each batch



# X-TED

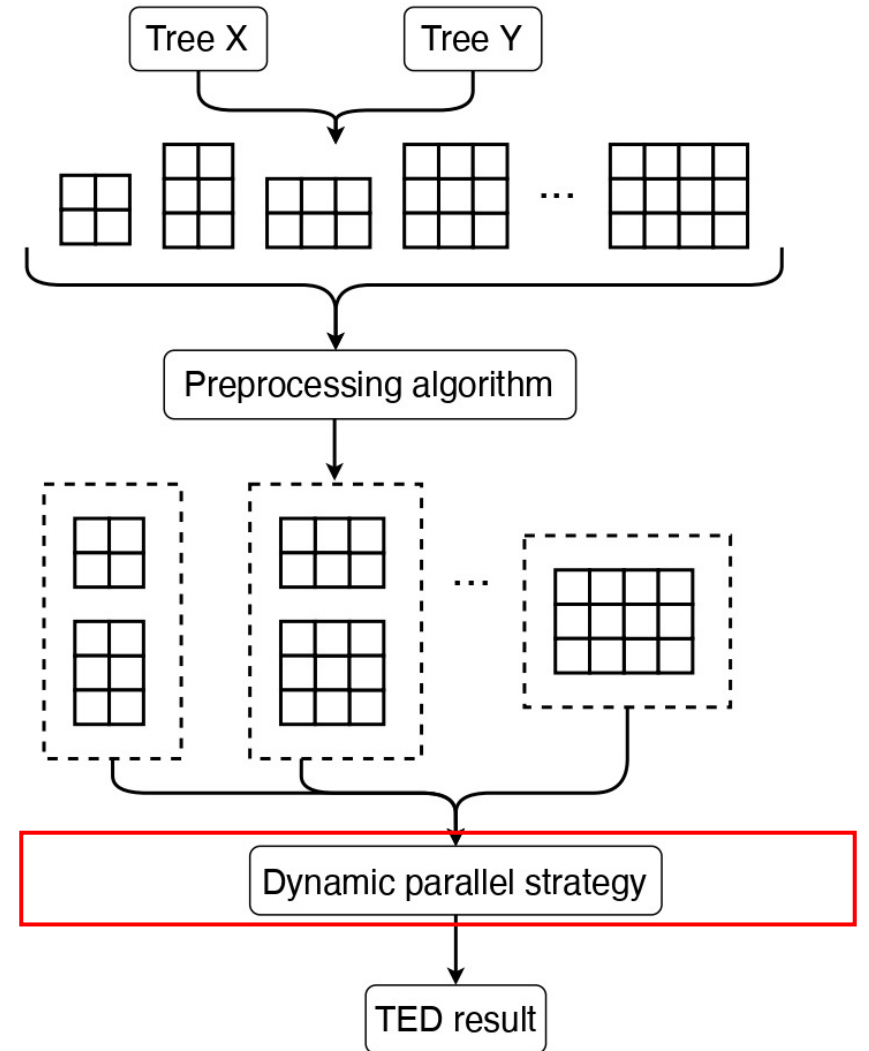
- Load-balanced





# X-TED

- Load-balanced
  - Different strategies for tables with different sizes



# Experiment Setup

<b>Dataset</b>	<b>Max. Depth</b>	<b>Avg. Depth</b>	<b>Avg. Nodes</b>	<b>Max. Nodes</b>
Swissport	9	7.01	988.36	7241
Python	156	13.11	927.41	8516
DBLP	7	3.16	26.05	1186
Bolzano	4	3.82	178.71	2105

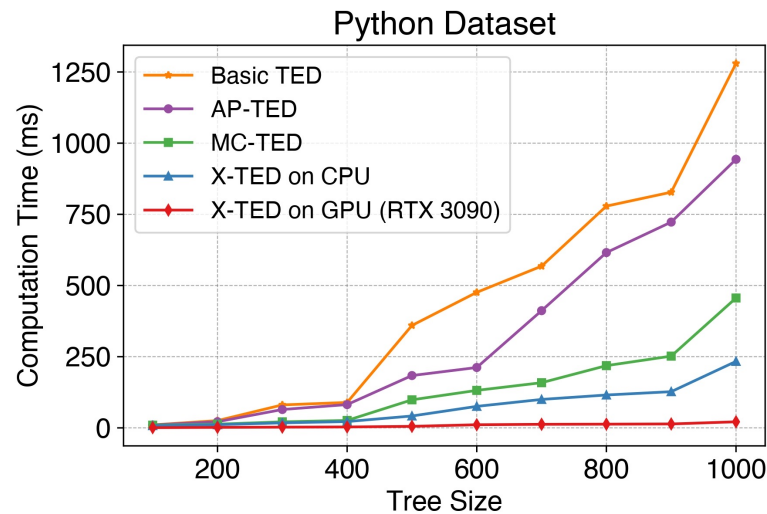
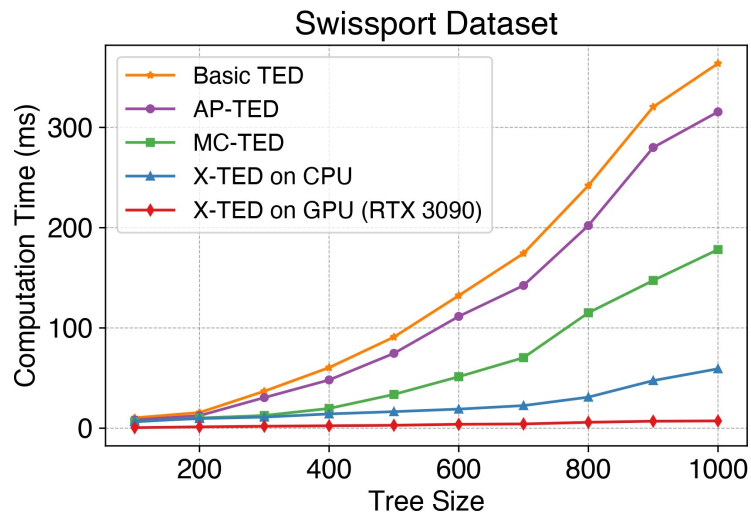
- Synthetic dataset: random recursive trees with 1000 to 9000 nodes
- For CPU baselines: Intel Core i9-12900 CPU, 8 Cores, 64GB
- For GPU baselines: NVIDIA RTX 3090, A100-SXM4, H100-PCIe

# High-Performance Results of X-TED

- 3 baselines:
  - Basic TED (1989)
  - State-of-the-art sequential solution (AP-TED, 2016)
  - State-of-the-art multi-core solution (MC-TED, 2020)

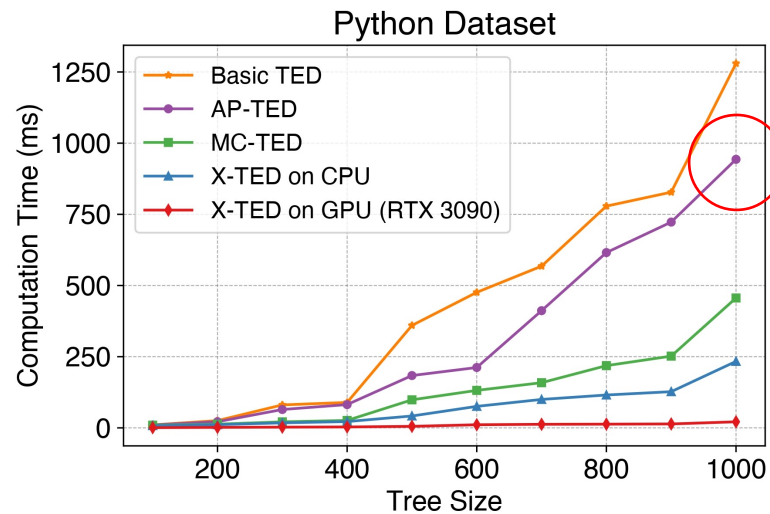
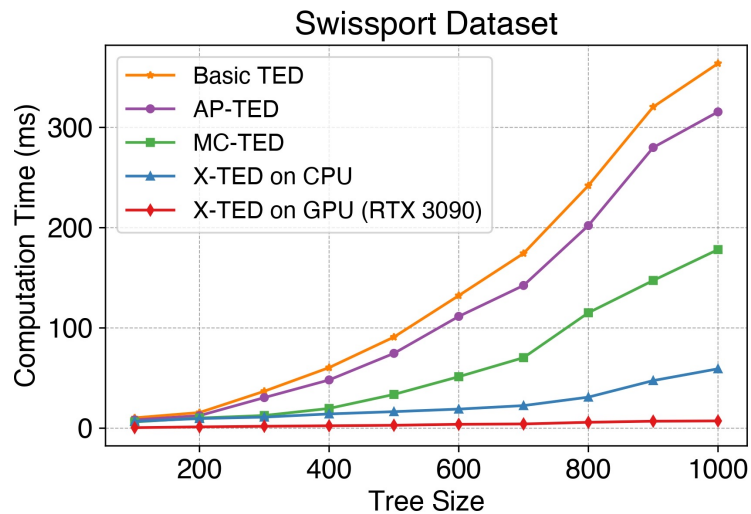
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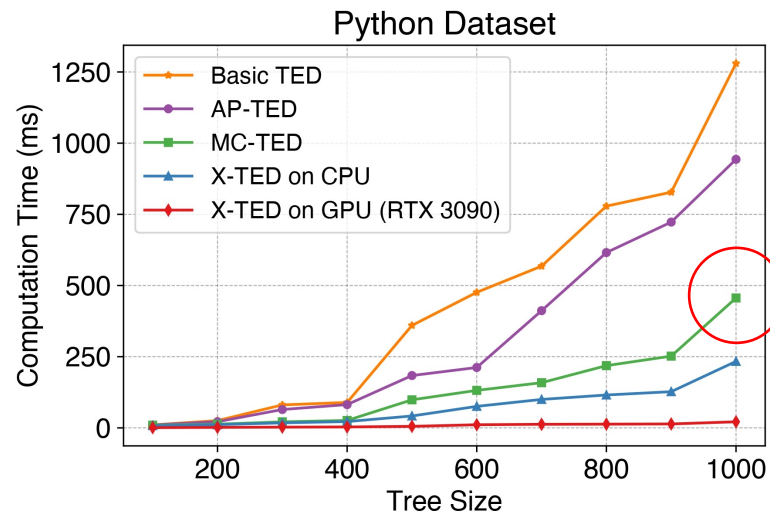
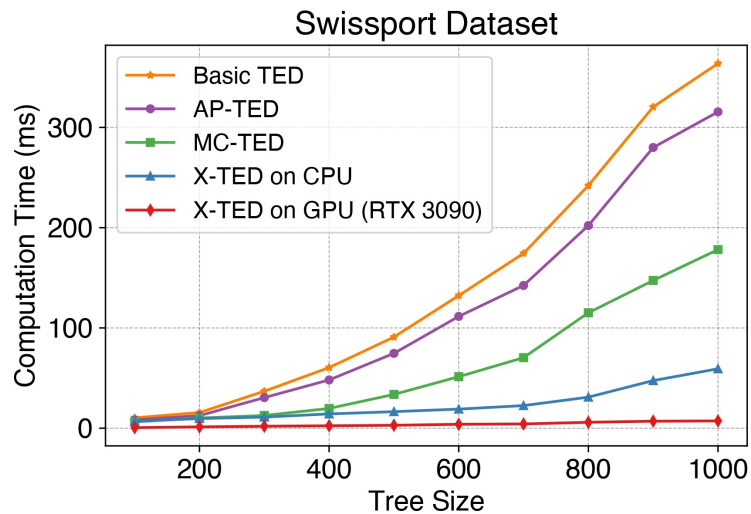
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- Speedup over AP-TED
  - X-TED (CPU): 4.8x
  - X-TED (GPU): 42x

# High-Performance Results of X-TED

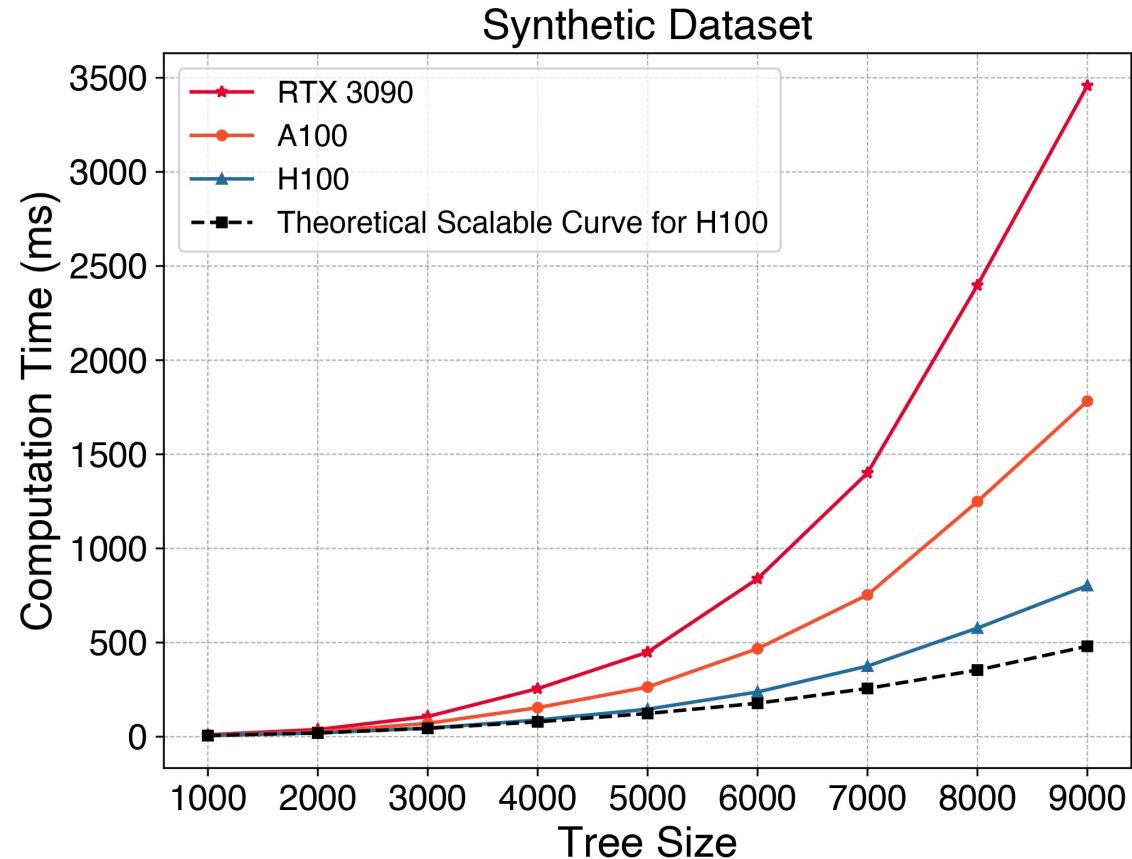
- 3 baselines:
  - Basic TED (1989)
  - State-of-the-art sequential solution (AP-TED, 2016)
  - State-of-the-art multi-core solution (MC-TED, 2020)



- Speedup over MC-TED
  - X-TED (CPU): 3.8x
  - X-TED (GPU): 31x

# High Scalability of X-TED

- Tree size: 1000 nodes → 9000 nodes



# Conclusion



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- X-TED
    - A massive parallel computation framework for TED
- The **best** parallel TED solution so far

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- Preprocessing ***enables*** massive TED parallel processing
- Dynamic parallel strategy ***adaptively*** utilizes GPU resources

Thank You!  
Email: fan.1090@osu.edu

\* Project Website and Open-Source Code: <https://github.com/Davis-Fan/X-TED>

# Image Reference

- Image in Page 13: Mohebbi, M., Razavi, S.N. & Balafar, M.A. Computing semantic similarity of texts based on deep graph learning with ability to use semantic role label information. Sci Rep 12, 14777 (2022). <https://doi.org/10.1038/s41598-022-19259-5>
- Image in Page 14: Chartier M, Chénard T, Barker J, Najmanovich R. Kinome Render: a stand-alone and web-accessible tool to annotate the human protein kinome tree. PeerJ. 2013 Aug 8;1:e126. doi: 10.7717/peerj.126. PMID: 23940838; PMCID: PMC3740139.
- Image in Page 15: Mate Kukri. 2022. Syntax searching C/C++ with Clang AST. Retrieved December 22, 2022 from <https://blog.trailofbits.com/2022/12/22/syntax-searching-c-c-clang-ast/>.
- Image in Page 16: Xutao Li, Yunming Ye, Mark Junjie Li, Michael K. Ng, On cluster tree for nested and multi-density data clustering, Pattern Recognition, Volume 43, Issue 9, 2010, Pages 3130-3143, ISSN 0031-3203, <https://doi.org/10.1016/j.patcog.2010.03.020>.