

A Hybrid Synchronization Mechanism for Parallel Sparse Triangular Solve

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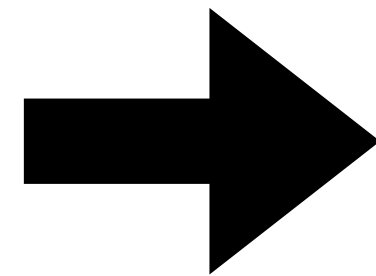


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Solving System of Linear Equations : An Example

Solve for \mathbf{y} in the equation $\mathbf{A}\mathbf{y} = \mathbf{b}$

$$\begin{aligned}y_1 + y_2 + y_3 &= 1 \\4y_1 + 3y_2 - y_3 &= 6 \\3y_1 + 5y_2 + 3y_3 &= 4\end{aligned}$$



$$\begin{matrix} & \mathbf{A} & & \mathbf{y} & & \mathbf{b} \\ & & & & & \\ \left(\begin{array}{ccc} 1 & 1 & 1 \\ 4 & 3 & -1 \\ 3 & 5 & 3 \end{array} \right) & & = & \left(\begin{array}{c} y_1 \\ y_2 \\ y_3 \end{array} \right) & & \left(\begin{array}{c} 1 \\ 6 \\ 4 \end{array} \right)\end{matrix}$$

Solving System of Linear Equations : An Example

To Solve $\mathbf{Ay} = \mathbf{b}$, Decompose $\mathbf{A} = \mathbf{LU}$

$$\begin{matrix} & \mathbf{A} & & & \mathbf{L} & & & \mathbf{U} \\ \left(\begin{array}{ccc} 1 & 1 & 1 \\ 4 & 3 & -1 \\ 3 & 5 & 3 \end{array} \right) & = & \left(\begin{array}{ccc} 1 & 0 & 0 \\ 4 & 1 & 0 \\ 3 & -2 & 1 \end{array} \right) & \left(\begin{array}{ccc} 1 & 1 & 1 \\ 0 & -1 & -5 \\ 0 & 0 & -10 \end{array} \right) \end{matrix}$$

Solving System of Linear Equations : An Example

To Solve $\mathbf{Ay} = \mathbf{b}$, Decompose $\mathbf{A} = \mathbf{LU}$ and solve $\mathbf{LUy} = \mathbf{b}$

$$\begin{pmatrix} 1 & 0 & 0 \\ 4 & 1 & 0 \\ 3 & -2 & 1 \end{pmatrix} \mathbf{L} \quad \begin{pmatrix} 1 & 1 & 1 \\ 0 & -1 & -5 \\ 0 & 0 & -10 \end{pmatrix} \mathbf{U} \quad \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} \mathbf{y} = \begin{pmatrix} 1 \\ 6 \\ 4 \end{pmatrix} \mathbf{b}$$

\mathbf{x}

Solve for \mathbf{x} in $\mathbf{Lx} = \mathbf{b}$, and then Solve for \mathbf{y} in $\mathbf{Uy} = \mathbf{x}$,
where L is a lower triangular matrix, and
U is an upper triangular matrix.

Dense Triangular Solve: Inherent Sequential Execution

Solve for \mathbf{x} in $\mathbf{Lx} = \mathbf{b}$

$$\begin{pmatrix} 1 & 0 & 0 \\ 4 & 1 & 0 \\ 3 & -2 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 1 \\ 6 \\ 4 \end{pmatrix}$$

Solving x_i
requires all the
values
from x_0 to x_{i-1}

$$\begin{array}{l} x_1 + 0 + 0 = 1 \\ 4x_1 + x_2 + 0 = 6 \\ 3x_1 - 2x_2 + x_3 = 4 \end{array} \quad \longrightarrow \quad \begin{array}{l} x_1 = 1 \\ x_2 = 6 - 4(1) = 2 \\ x_3 = 4 - 3(1) + 2(2) = 5 \end{array}$$

Sparse Triangular Solve (SpTS)

Solve for \mathbf{x} in the equation $\mathbf{Lx} = \mathbf{b}$

0	1							
1	1	1						
2	1		1					
3		1		1				
4		1		1	1			
5	1	1	1			1		
6						1	1	
7				1				1

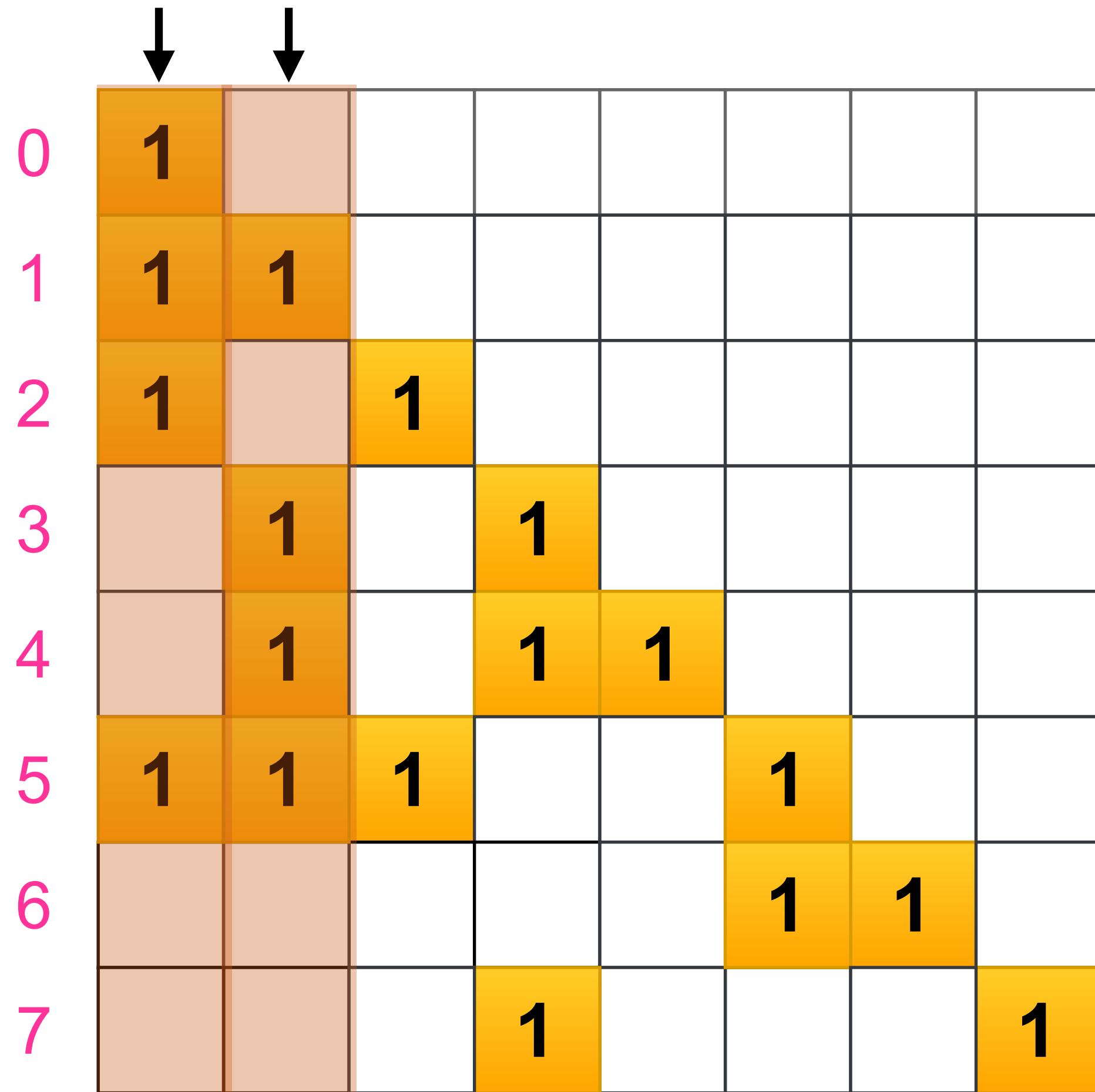
Lower Triangular Sparse Matrix

$$\begin{matrix} * \\ = \end{matrix} \begin{matrix} x_0 \\ x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \end{matrix} = \begin{matrix} b_0 \\ b_1 \\ b_2 \\ b_3 \\ b_4 \\ b_5 \\ b_6 \\ b_7 \end{matrix}$$

Solving x_i may not require all the values from x_0 to x_{i-1}

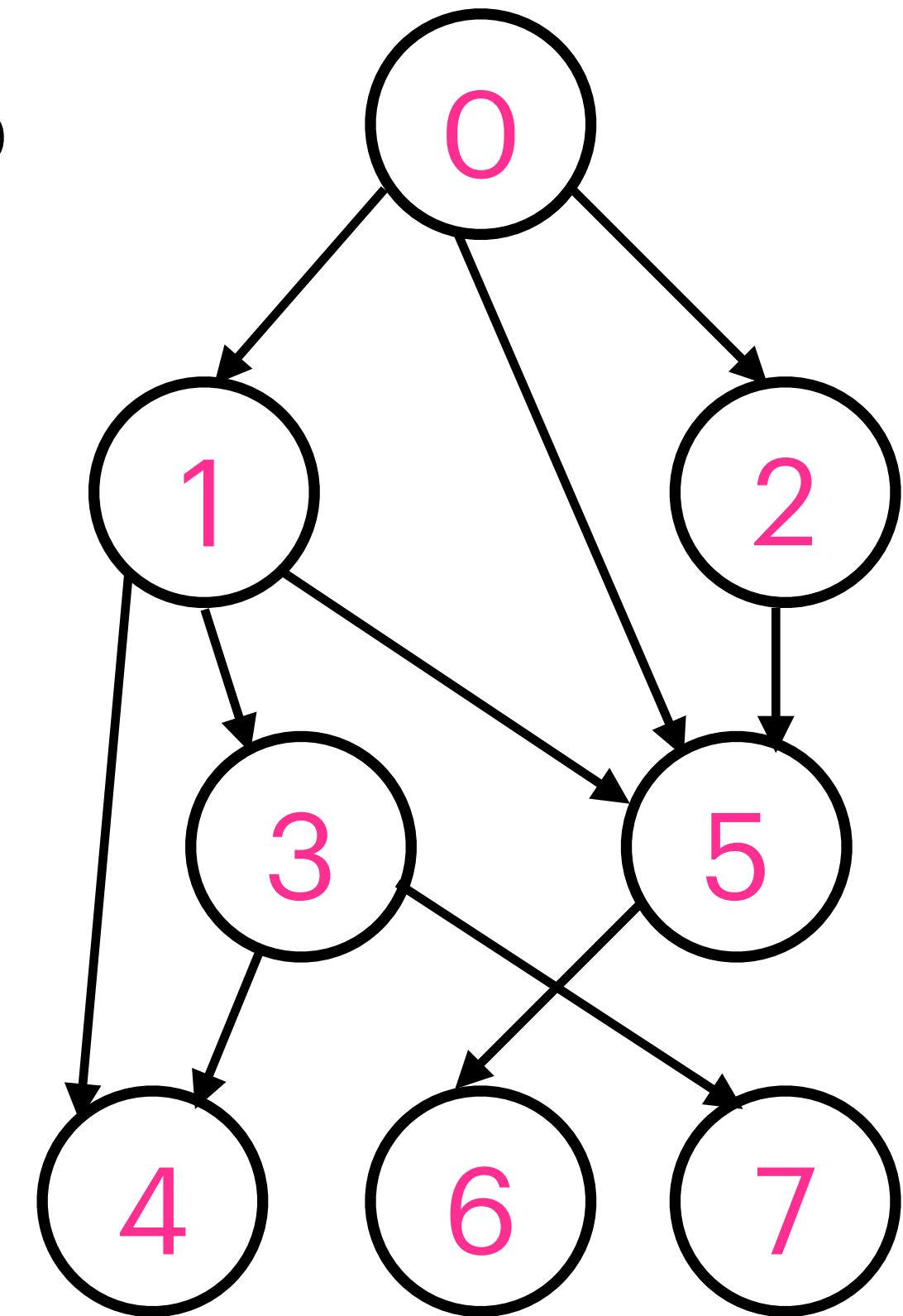
Sparse Triangular Solve (SpTS) : Task Dependency Graph

Solve for \mathbf{x} in the equation $\mathbf{Lx} = \mathbf{b}$



Lower Triangular Sparse Matrix

$$\begin{matrix} * \\ \end{matrix}
 \begin{matrix} x_0 \\ x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \end{matrix}
 =
 \begin{matrix} b_0 \\ b_1 \\ b_2 \\ b_3 \\ b_4 \\ b_5 \\ b_6 \\ b_7 \end{matrix}$$



Therefore, SpTS has the potential to be computed in parallel.

Parallel SpTS : Existing Synchronization Methods

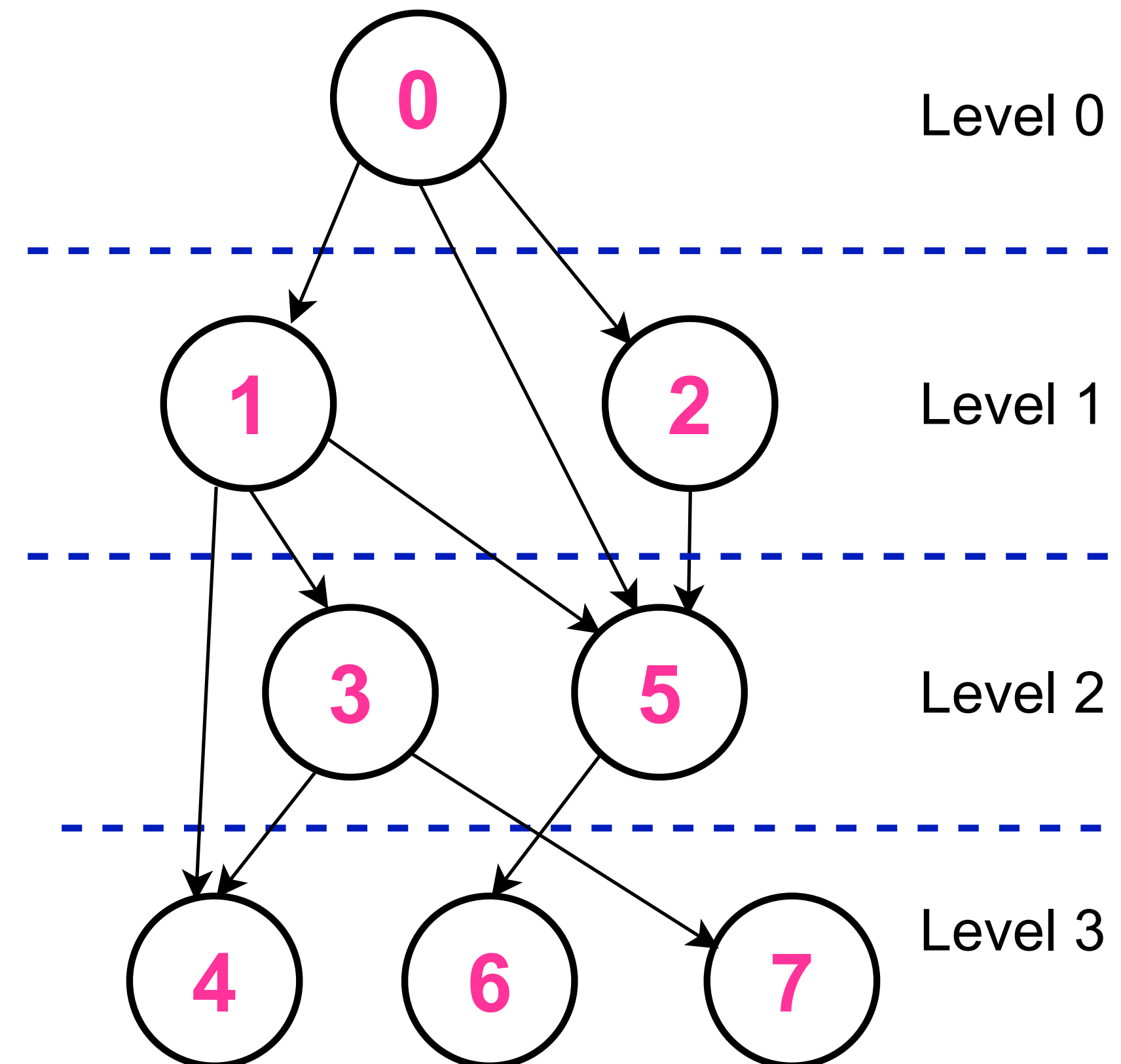
Parallel SpTS: Level-set Method

Solve for \mathbf{x} in the equation $\mathbf{Lx} = \mathbf{b}$

- Make sets of the matrix-rows which can be solved **independently** and **simultaneously**.
- Dependency graph represents the level-set formation.
- Uses barrier synchronization.

0	1						
1	1	1					
2	1		1				
3		1		1			
4		1		1	1		
5	1	1	1			1	
6						1	1
7				1			1

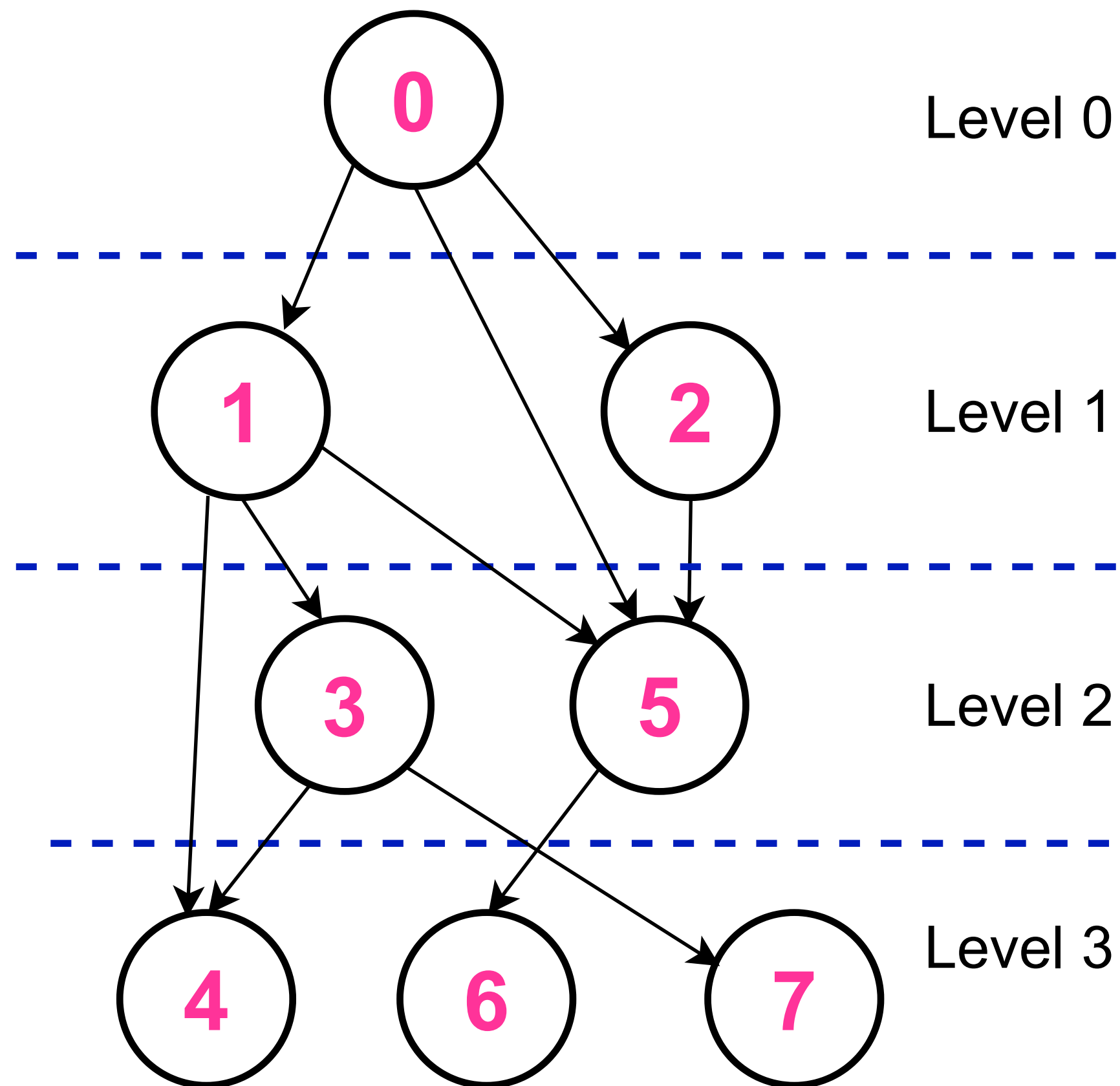
Level-set formation:



Parallel SpTS : Level-set Method

Solve for \mathbf{x} in the equation $\mathbf{Lx} = \mathbf{b}$

Level-set formation:



Works well when :

- Balanced workload among the workers at each level.
- A small number of levels

0	1						
1	1	1					
2	1		1				
3		1		1			
4		1		1	1		
5	1	1	1			1	
6						1	1
7				1			1

```
for each level
  for each row i inside the level in parallel
    Solve x[i]
  end for
// barrier synchronization
end for
```

Parallel SpTS : Synchronization-free Method

Solve for \mathbf{x} in the equation $\mathbf{Lx} = \mathbf{b}$

- Eliminate the pre-processing step.
- Uses atomic operations for busy-waiting.
- Effective for GPUs.

```
for each row i in parallel
  for each dependent row j
    while atomic_read(flag[j]) != 1
      // busy-wait
    end while
    Solve x[i]
  end for
  Solve x[i]
  atomic_write(flag[i], 1)
end for
```

0	1						
1	1	1					
2	1		1				
3		1		1			
4		1		1	1		
5	1	1	1			1	
6						1	1
7				1			1

Limitations of the existing methods

Level-set method

- Large number of level-sets -> costly barrier synchronization.
- Small and varied number of components per level -> waste the assigned CPU resources.
- Uneven distribution of non-zeros among the rows -> load imbalance.

Synchronization-free method

- Highly impractical for CPUs due to the heavy use of expensive atomic and busy-waiting operations on the limited number of threads.

Our Objective

Improve the performance of parallel SpTS for WebAssembly on CPUs

How?

- 1. Avoid synchronization barriers !**
- 2. Minimize the use of atomic operations as much as possible !**

Why WebAssembly?

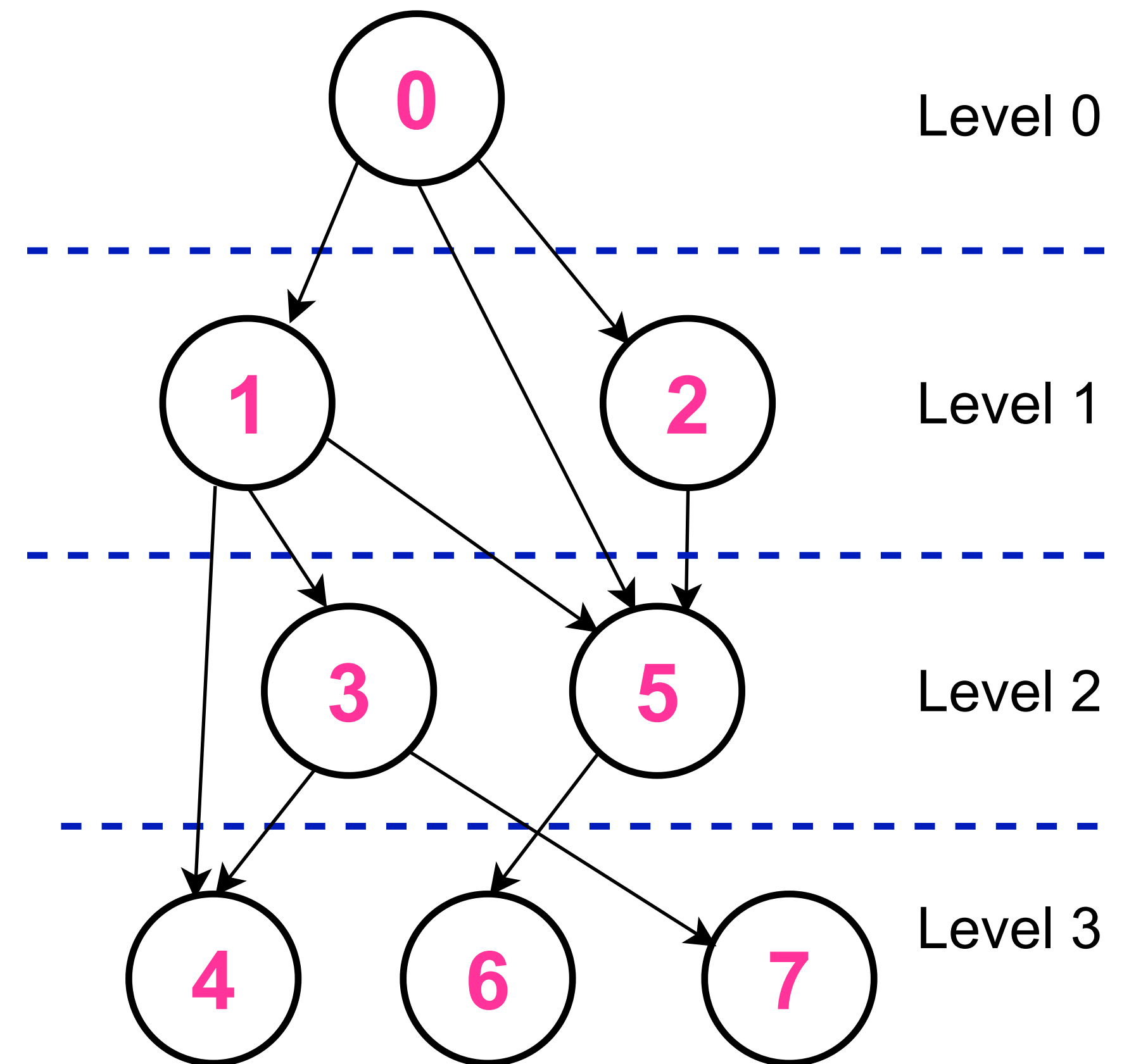
- 1. A new low-level target language for the web.**
- 2. Building efficient web-based sparse matrix kernels for ML.**

No synchronization barriers

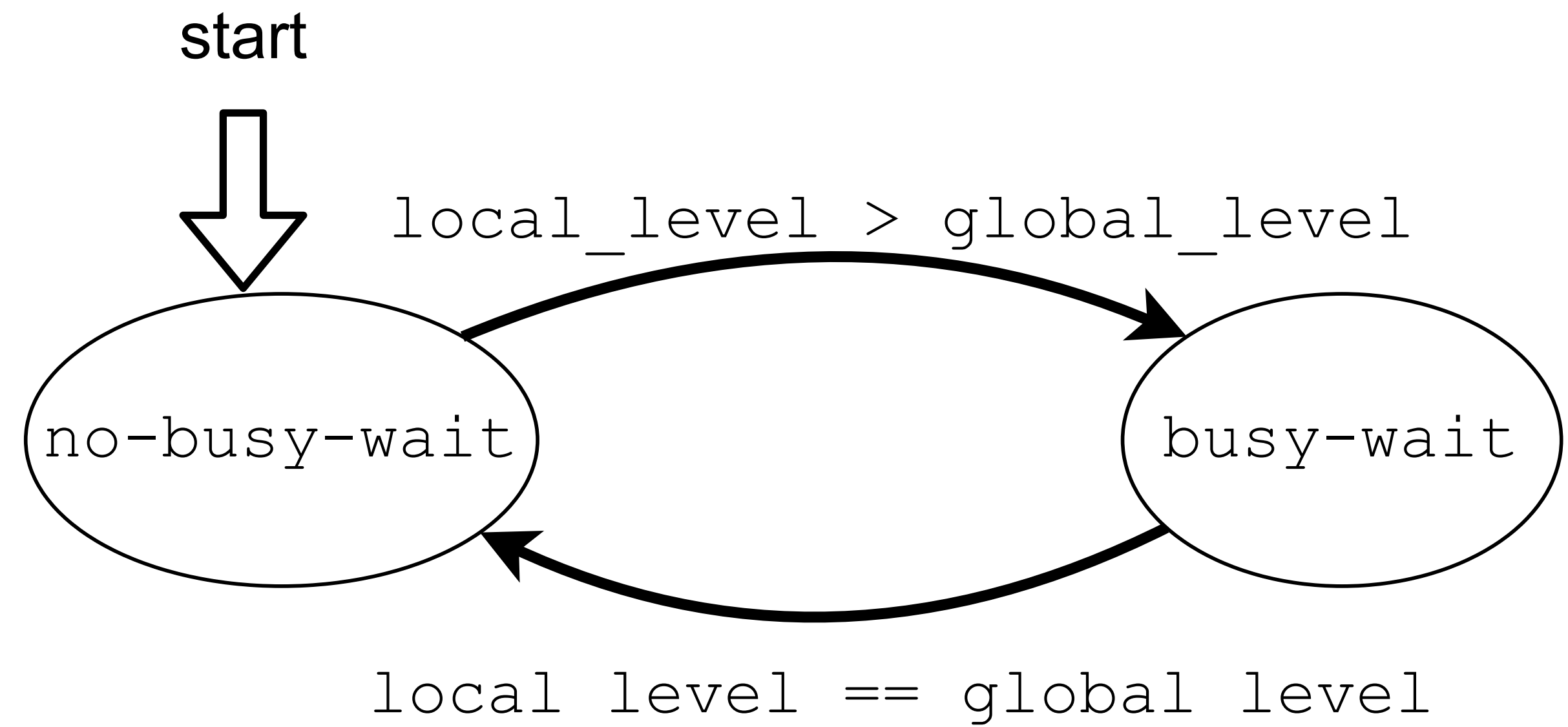
But with level-set formation

- Keep the pre-processing step of level-set method.
- Why? A systematic way to guarantee: the threads at the same level can make progress independently and simultaneously.
- Spatial locality benefits from the level-set formation.
- Additionally, allow each thread to immediately process the next level after the completion of its work at the current level.

Level-set formation:



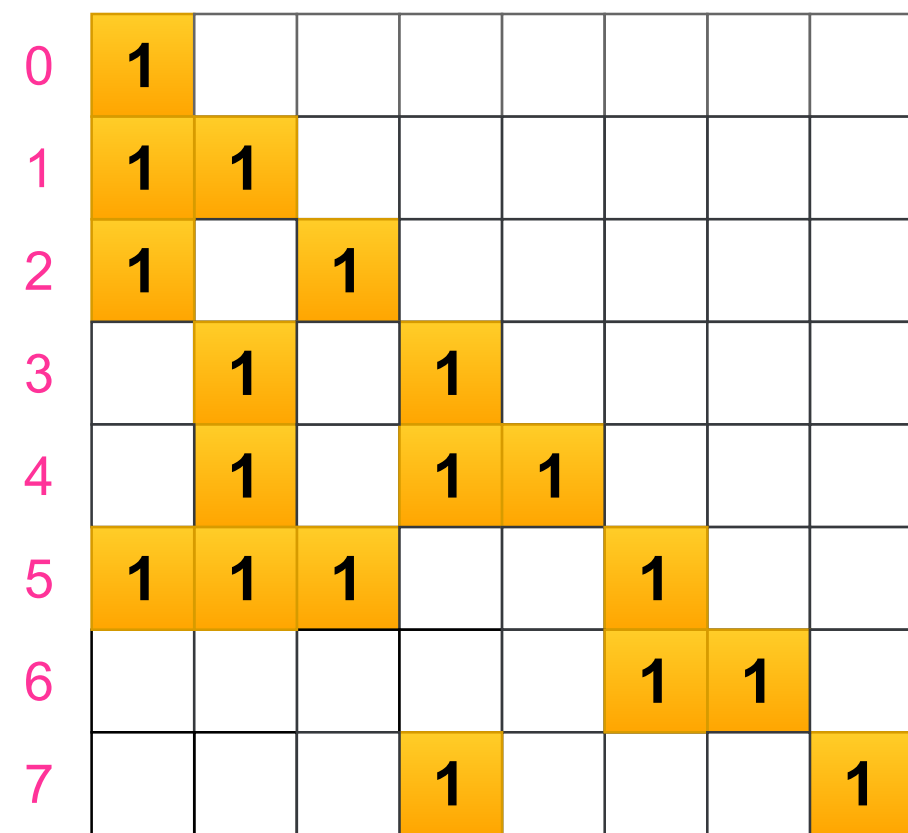
Our Technique: Two Synchronization Modes



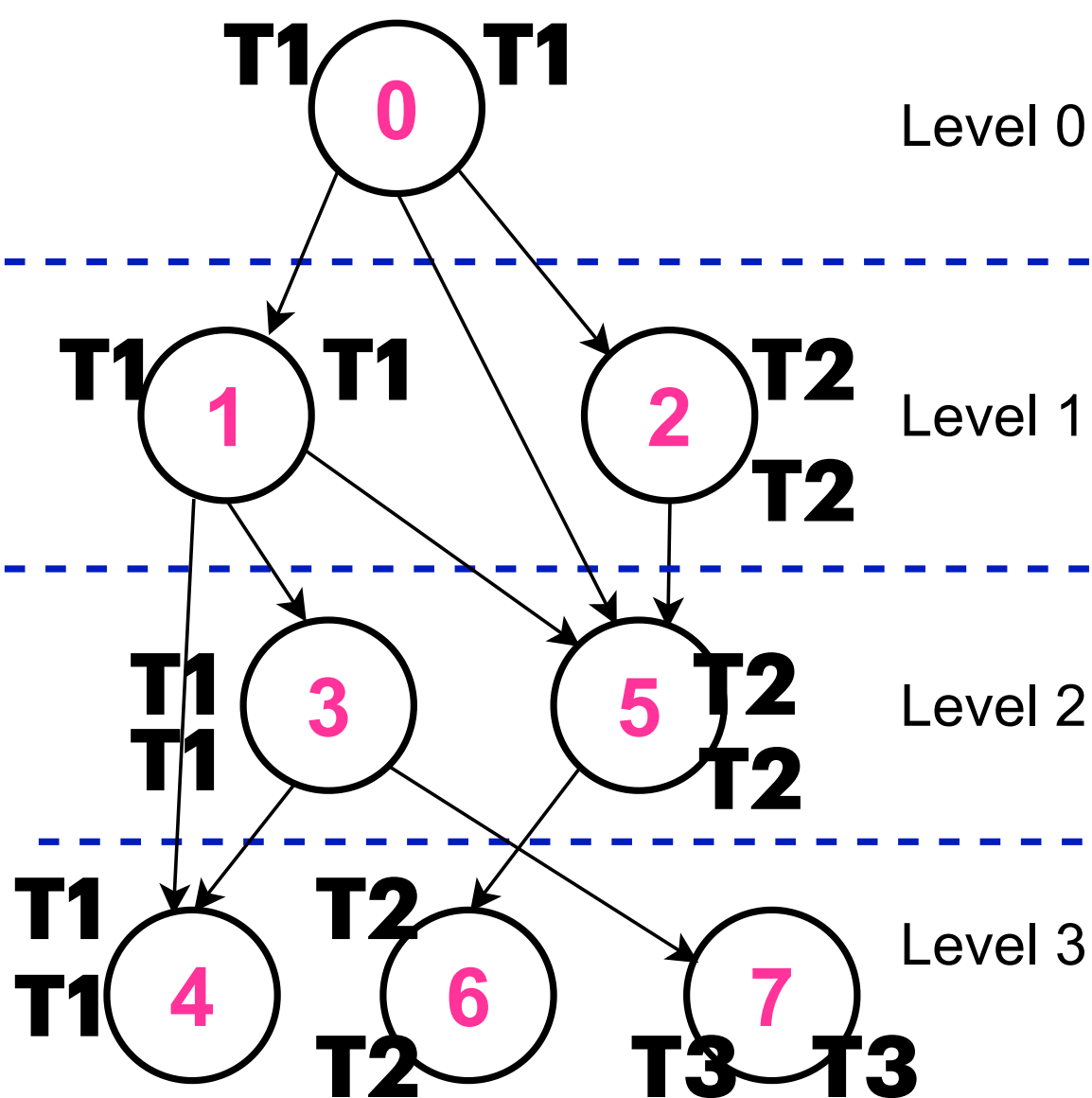
- **no-busy-wait (default)** : when the *current working level of the thread* (**local_level**) is equal to the *maximum working level achieved by all the threads* (**global_level**).
- **busy-wait** : when *local_level* is greater than *global_level*, indicating that the thread is presently working in the advanced levels.
- Each thread can dynamically switch between the two modes as many times as required.

An example to contrast the parallel SpTS workflow

A tuple (**current_level**, **row**, **column**) represents the values of these parameters for each thread.



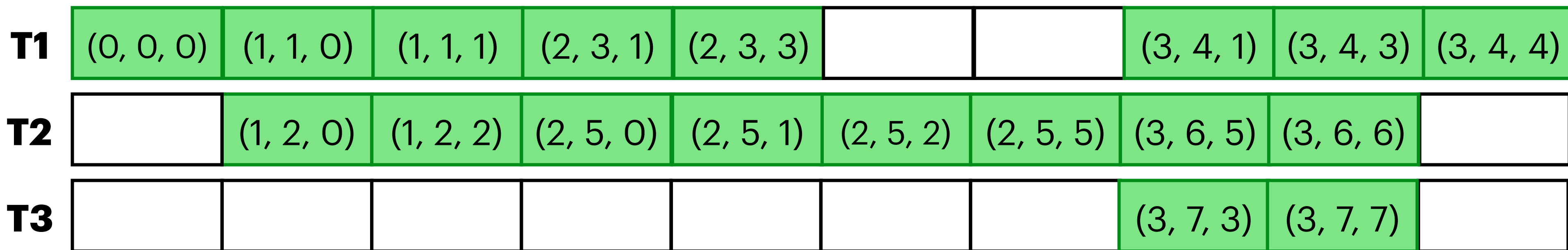
Level-set formation:



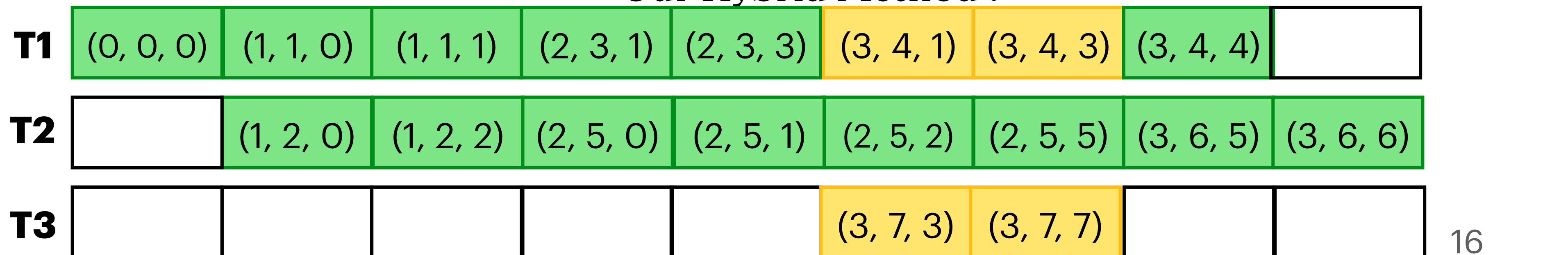
idle thread
 no-busy-wait
 busy-wait



Level-set Method :



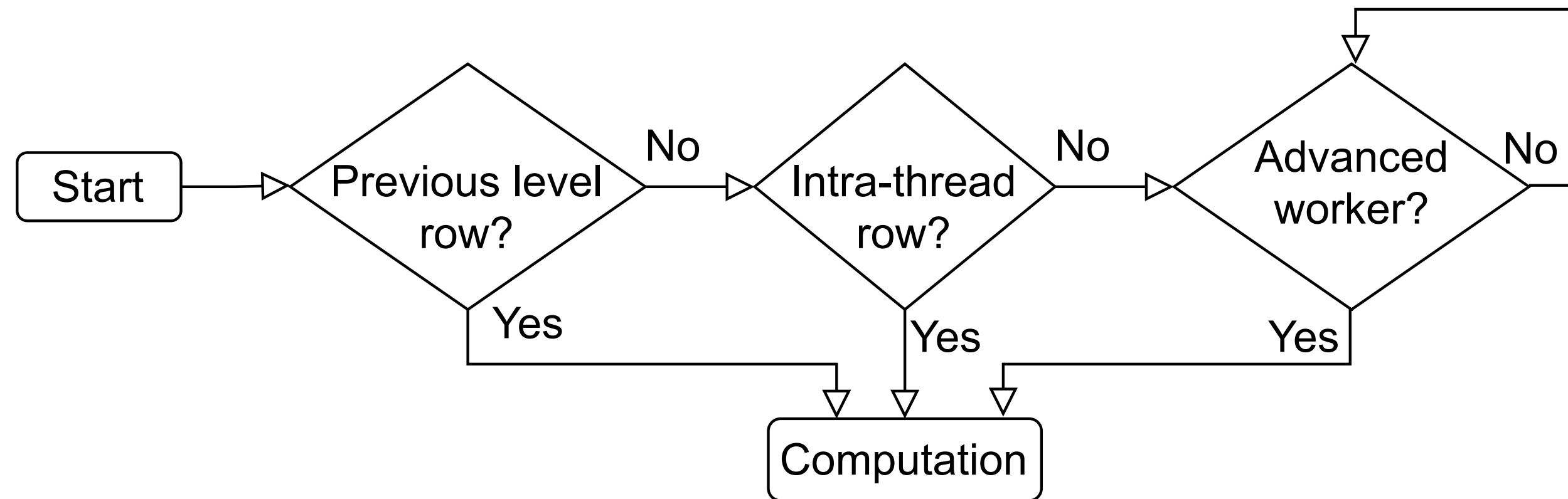
Our Hybrid Method :



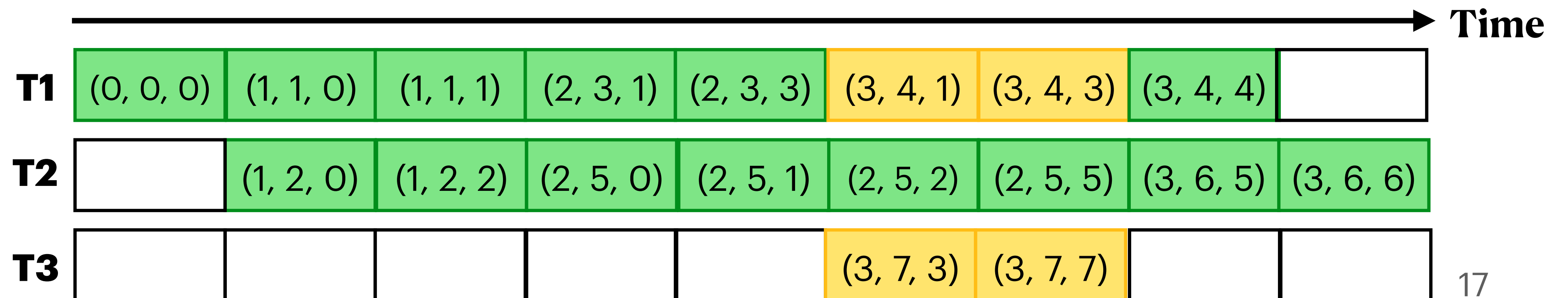
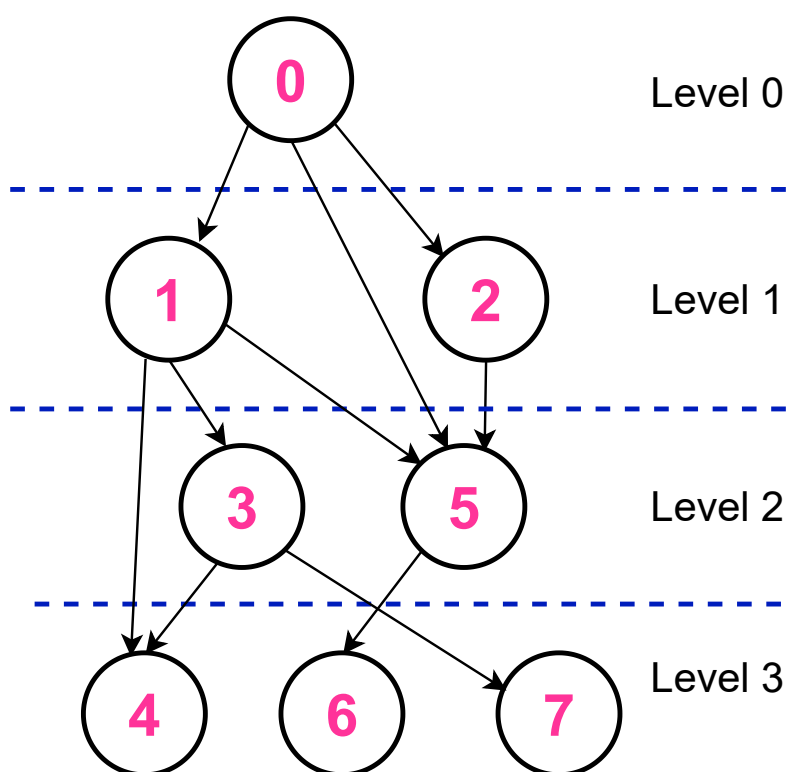
busy-wait synchronization mode

Classify required-rows into 4 exhaustive categories : allows progress

- Previous-level rows
- Intra-thread rows
- Advanced-worker inter-thread rows
- Inter-thread rows



Level-set formation:



SpTS Performance Comparison

Our Hybrid Method vs Level-set Method : higher is better

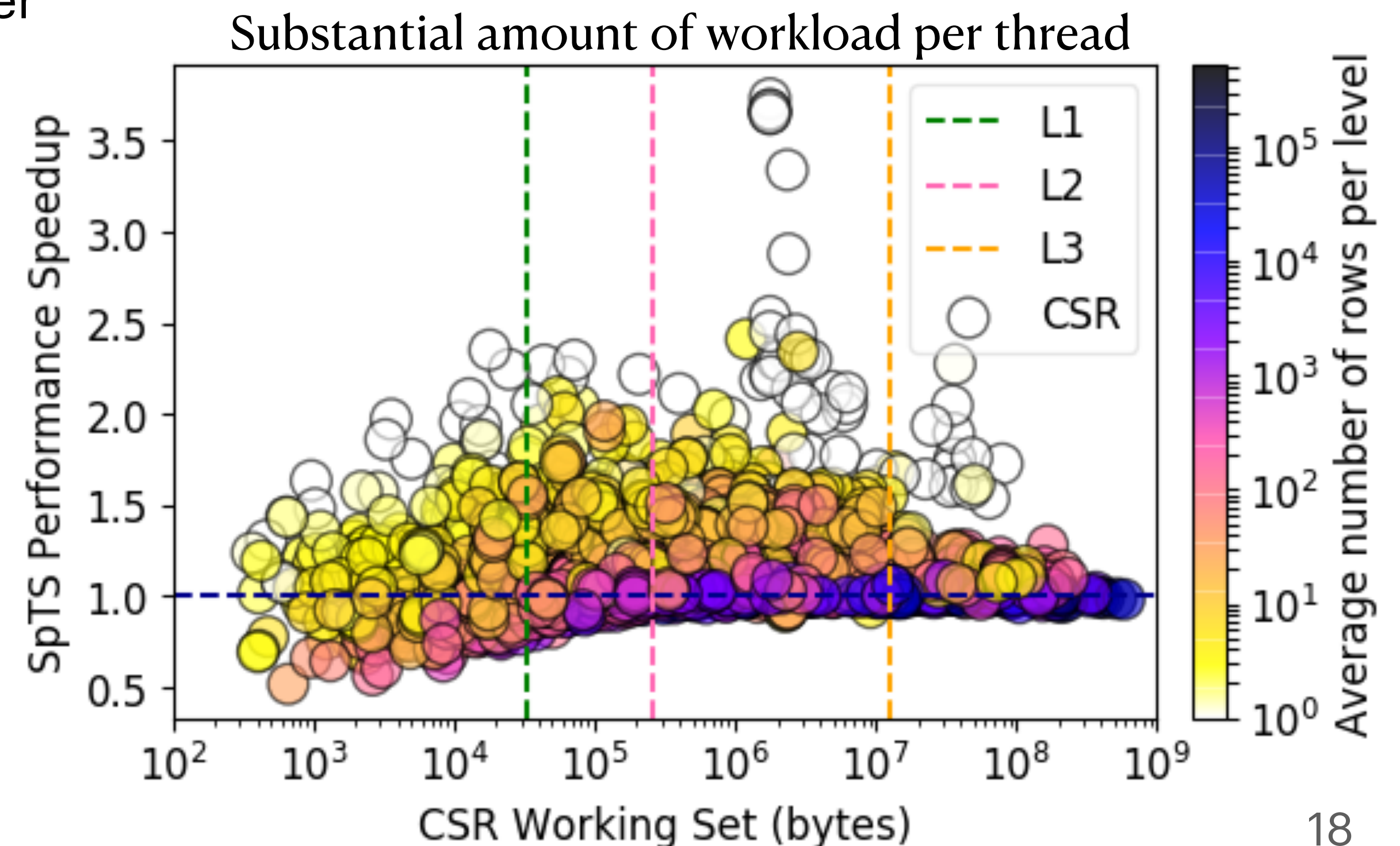
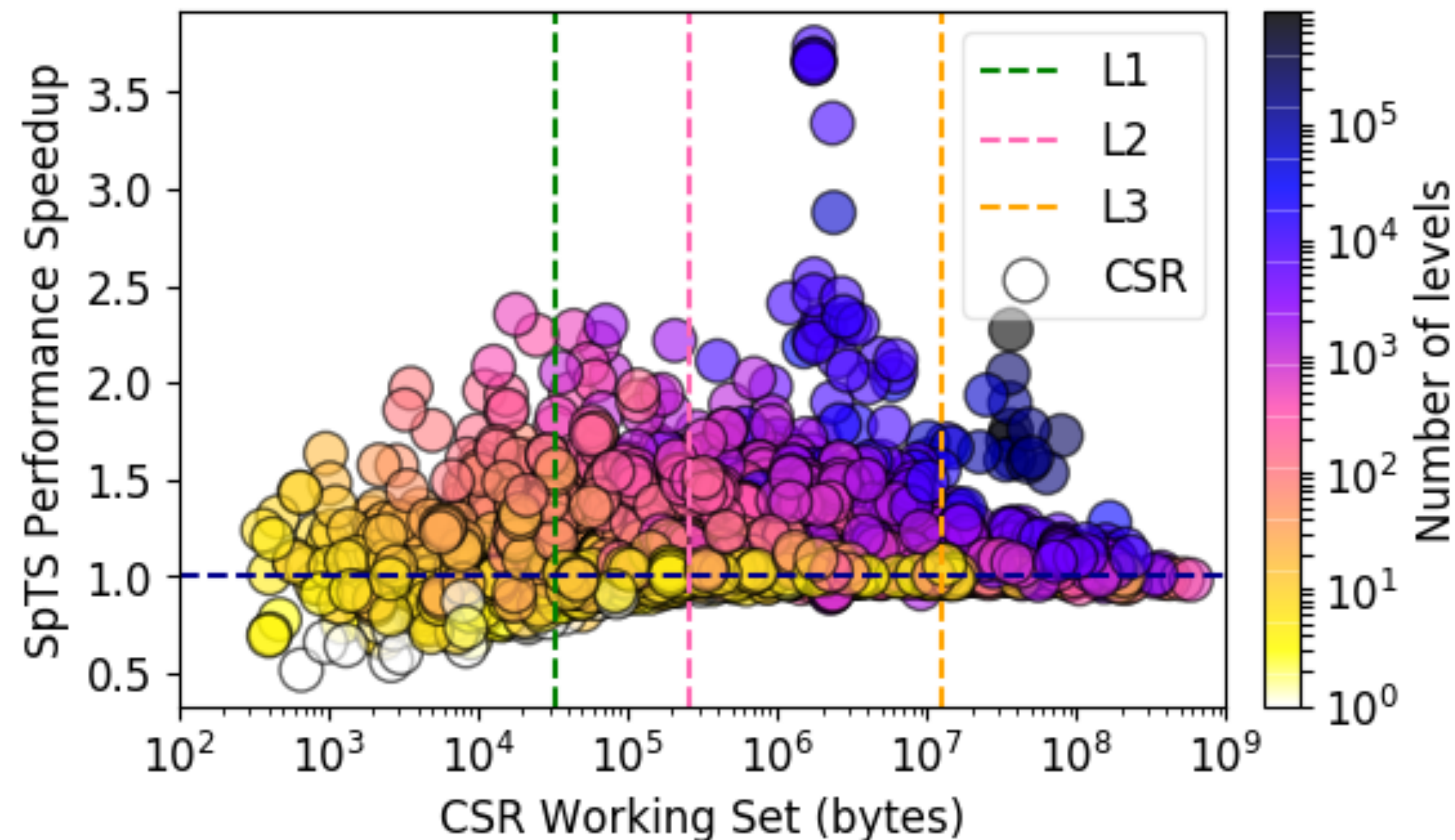
Machine : Intel Core i7-3930K with 6 3.20GHz cores, 12MB last-level cache and 16GB memory, running Ubuntu Linux 18.04.5

Input : 1957 real-life sparse matrices from The SuiteSparse Matrix Collection.

Storage Format : CSR (Compressed Sparse Row)

Target Language : WebAssembly & JavaScript

Execution Environment : Chrome 92 headless browser



Below 1

Matrices where level-set method performs better than hybrid method

Matrix	N	nnz	nlevels	N/nlevels	Level-set Performance (GFLOPS)	Hybrid Performance (GFLOPS)	Speedup
t2dal_e	4257	4257	1	4257	1.70	1.52	0.89x
bcpwr08	1624	3837	14	116	0.89	0.74	0.83x
t3dl_a	20360	265113	633	32.2	1.32	1.22	0.92x
exdata_1	6001	1137751	1501	3.99	1.47	1.37	0.93x

NS : not shown in the table, investigated separately

- Large number of rows per level -> substantial amount of workload for each thread.
- Nearly balanced workload among the threads at each level (NS) -> insignificant cost of barrier synchronization.
- Small number of rows per level with large number of non-zeros per row (NS) -> minimal parallelism but likely reduced the cost of barrier synchronization.

Above 1

Matrices where hybrid method performs better than level-set method

Matrix	N	nnz	nlevels	N/nlevels	Level-set Performance (GFLOPS)	Hybrid Performance (GFLOPS)	Speedup
lung2	109460	273647	479	228.5	1.47	2.28	1.55x
delaunay_n17	131072	524248	910	144	1.36	1.82	1.34x
e40r0100	17281	257727	512	33.7	1.49	2.06	1.38x
smt	25710	1887646	4646	5.5	1.27	1.72	1.35x

NS : not shown in the table, investigated separately

- Large number of levels -> increased the cost of barrier synchronization for level-set method.
- Small to moderate number of rows per level -> limited amount of workload for each thread.
- Uneven distribution of rows among the levels (NS) -> limits the amount of workload per thread and waste CPU resources at the barriers.
- Hybrid method benefits by allowing the threads to move to further levels to perform some feasible computation.²⁰

Close to 1

Matrices where hybrid method performs similar to level-set method

Matrix	N	nnz	nlevels	N/nlevels	Level-set Performance (GFLOPS)	Hybrid Performance (GFLOPS)	Speedup
t3dl_e	20360	20360	1	20360	1.87	1.83	0.98x
mbeacxc	496	30309	214	2.3	0.76	0.76	1.00x
coPapersCiteseer	434102	16470822	8087	53.7	2.26	2.24	0.99x
kron-g500-logn18	262144	10844830	1820	144	1.21	1.19	0.98x

NS : not shown in the table, investigated separately

- Presence of diagonal matrices in both Below 1 and Close to 1 categories -> overhead of our method becomes insignificant for the large matrices with small number of levels.
- Large number of levels with little imbalanced workload (NS) -> overhead cancels out the performance gain.

Summary

- We employ the level-set formation without barrier synchronization, and make minimal use of expensive atomic operations by dynamically switching between the two synchronization modes as required.
- We evaluate the performance of hybrid method over level-set method using our WebAssembly implementations on around 2000 sparse matrices.
- Our evaluations show the potential of our method to support the adaptive synchronization techniques in the future.

Future Directions

- Explore more sparse storage formats and apply optimization techniques like SIMD.
- Employ the upcoming synchronization constructs like floating-point atomics from the rapidly evolving WebAssembly instruction set.
- Investigate pertinent matrix structure features to develop an adaptive synchronization method (I mean build a “sorting hat”!) in the future (and perhaps call our strategy to be “Ravenclaw”).

Contact us :

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Webpage : www.cs.mcgill.ca/~psandh3

Extras

busy-wait synchronization mode

```
(f32.load (local.get $csr_val))
(i32.load (local.get $csr_col))
(local.set $required_row)
(i32.atomic.load (local.get $global_row_index))
(local.get $required_row)
(i32.le_s)
if
  (i32.load (i32.add (local.get $row_worker_index) (i32.shl (local.get $required_row)
    (i32.const 2))))
  (local.set $worker)
  (local.get $worker)
  (local.get $current_worker)
  (i32.ne)
  if
    (i32.load (i32.add (local.get $row_level_index) (i32.shl (local.get $required_row)
      ) (i32.const 2))))
    (local.set $required_level)
    (loop $busy_wait_loop
      (local.get $required_level)
      (i32.atomic.load (i32.add (local.get $worker_level_index) (i32.shl (local.get
        $worker) (i32.const 2))))
      (i32.gt_s)
      (br_if $busy_wait_loop)
    )
  )
end
end
(f32.load (i32.add (local.get $x) (i32.shl (local.get $required_row) (i32.const 2))))
(f32.mul)
```