

# Parallel Prefix Scan

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Some slides/material from:  
UIUC course by Wen-Mei Hwu and David Kirk  
IISC-SERC course by Mike Giles, Oxford University Mathematical Institute

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## Parallel Prefix Scan: Why Do We Care?

- Parallel Prefix Sum (Scan) algorithms
  - One of the most frequently-used parallel patterns
  - frequently used for parallel work assignment and resource allocation
  - A key primitive in many parallel algorithms to convert serial computation into parallel computation
  - Based on reduction tree and reverse reduction tree
- Reading – Mark Harris, Parallel Prefix Sum with CUDA
  - [http://developer.download.nvidia.com/compute/cuda/1\\_1/Website/projects/scan/doc/scan.pdf](http://developer.download.nvidia.com/compute/cuda/1_1/Website/projects/scan/doc/scan.pdf)

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## (Inclusive) Prefix-Sum (Scan) Definition

**Definition:** The all-prefix-sums operation takes a binary associative operator  $\oplus$ , and an array of  $n$  elements

$$[x_0, x_1, \dots, x_{n-1}],$$

and returns the array

$$[x_0, (x_0 \oplus x_1), \dots, (x_0 \oplus x_1 \oplus \dots \oplus x_{n-1})].$$

**Example:** If  $\oplus$  is addition, then the all-prefix-sums operation on the array  $[3 \ 1 \ 7 \ 0 \ 4 \ 1 \ 6 \ 3]$ , would return  $[3 \ 4 \ 11 \ 11 \ 15 \ 16 \ 22 \ 25]$ .

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## A Inclusive Scan Application Example



World's longest sausage (38.99 miles)

World's longest hot dog (60m)



- Assume a 100-inch sausage to feed 10
- We know how much each person wants in inches
  - $[3 \ 5 \ 2 \ 7 \ 28 \ 4 \ 3 \ 0 \ 8 \ 1]$
- How do we cut the sausage quickly?
- How much will be left?
- **Method 1:** cut the sections sequentially: 3 inches first, 5 inches second, 2 inches third, etc.
- **Method 2:** calculate Prefix Scan
  - $[3, 8, 10, 17, 45, 49, 52, 52, 60, 61]$  (39 inches left)

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## Other Applications

- Allocating memory to parallel threads
  - Think of supercomputers with 1-million-thread workloads
  - Not science fiction:
    - Sunway TaihuLight @ NSC, China, 2016: 10,649,600 SW26010 cores
    - Titan @ ORNL, 2012: 299,008 x86 cores + 50,233,344 CUDA cores
    - Summit @ ORNL, 2018: 221,184 Power9 cores + 141,557,760 CUDA cores
- Allocating memory buffer to communication channels
- ...

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## An Inclusive Sequential Prefix-Sum

Given a sequence  $[x_0, x_1, x_2, \dots]$

Calculate output  $[y_0, y_1, y_2, \dots]$

Such that

$$y_0 = x_0$$
$$y_1 = x_0 + x_1$$
$$y_2 = x_0 + x_1 + x_2$$

...

*Using a recursive definition*

$$y_i = y_{i-1} + x_i$$

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### A Work-Efficient C Implementation

```
y[0] = x[0];  
for (i = 1; i < Max_i; i++)  
    y[i] = y[i-1] + x[i];
```

Computationally efficient:

N additions needed for N elements -  $O(N)$ !

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### A Naïve Inclusive Parallel Scan

- Assign one thread to calculate each y element
- Have every thread to add up all x elements needed for the y element

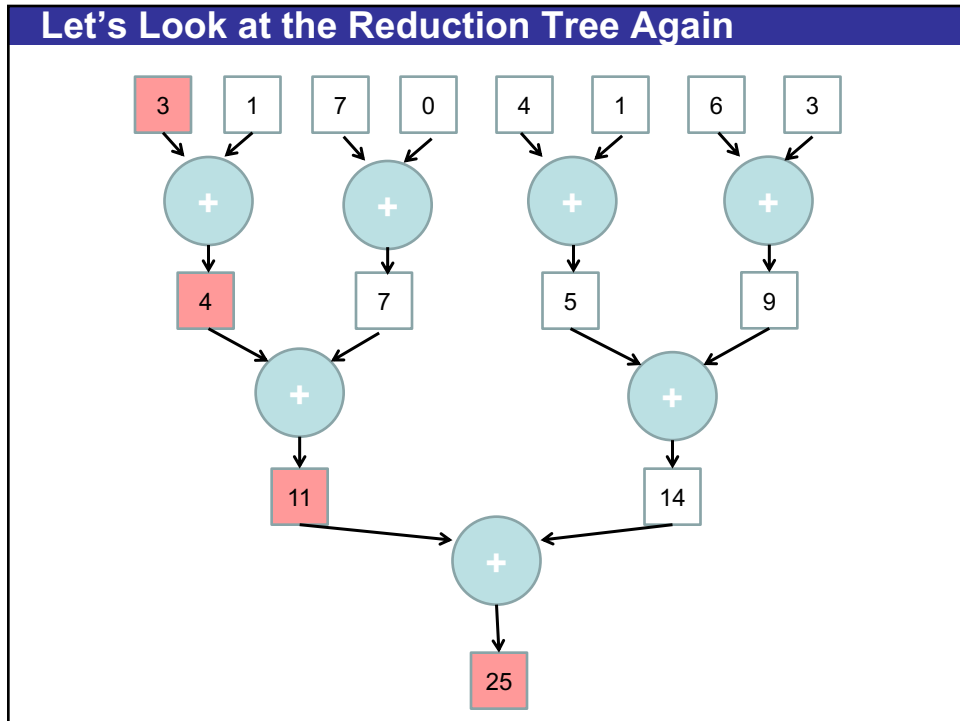
$$y_0 = x_0$$

$$y_1 = x_0 + x_1$$

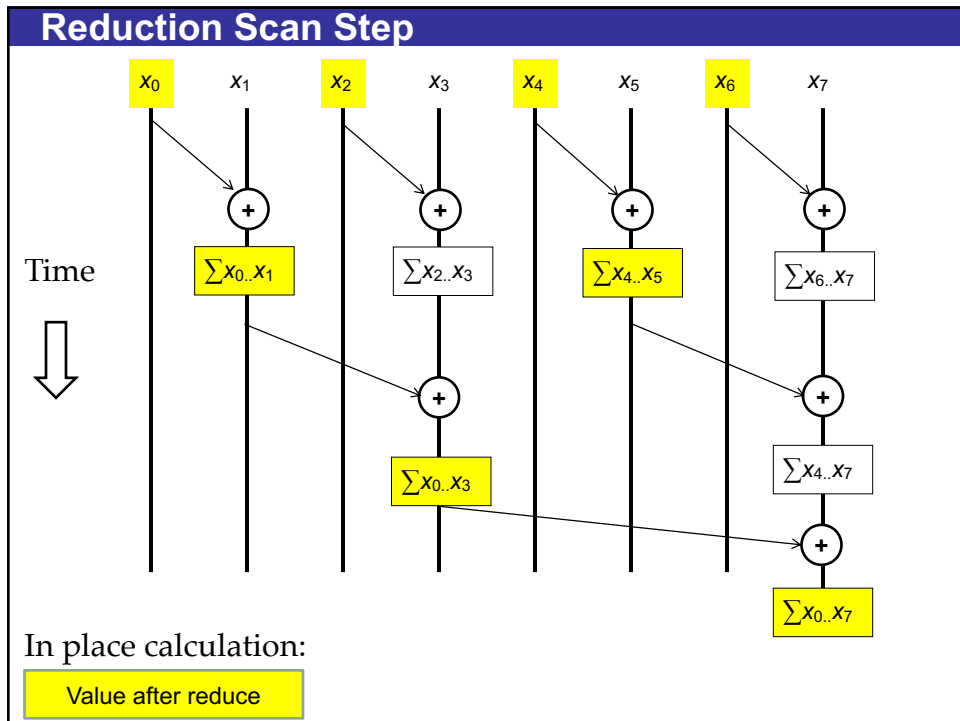
$$y_2 = x_0 + x_1 + x_2$$

“Parallel programming is easy as long as you do not care about performance.”

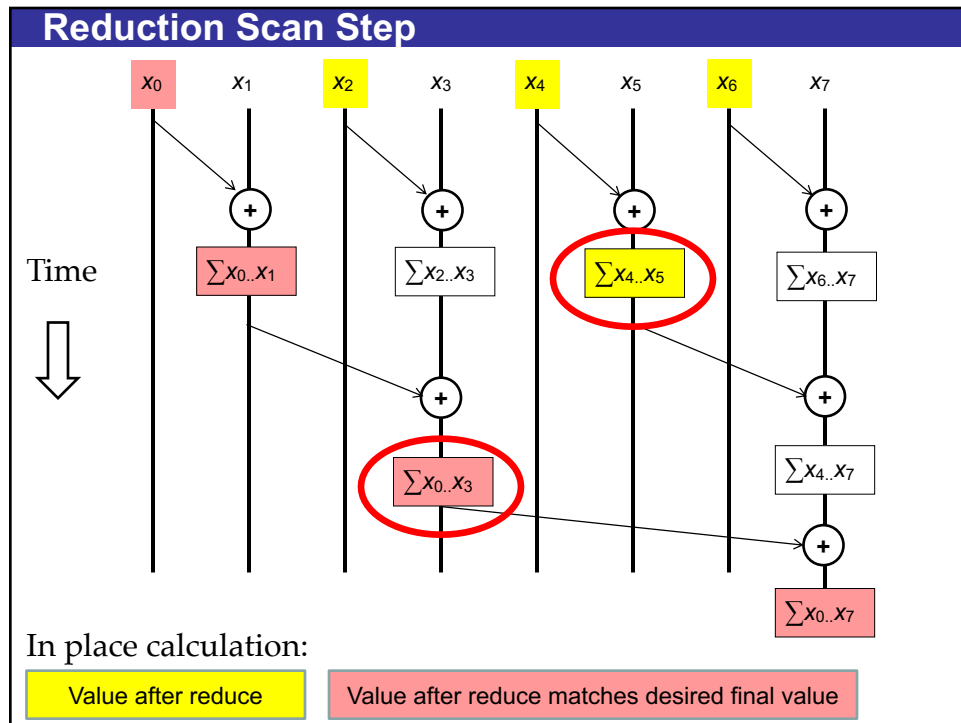
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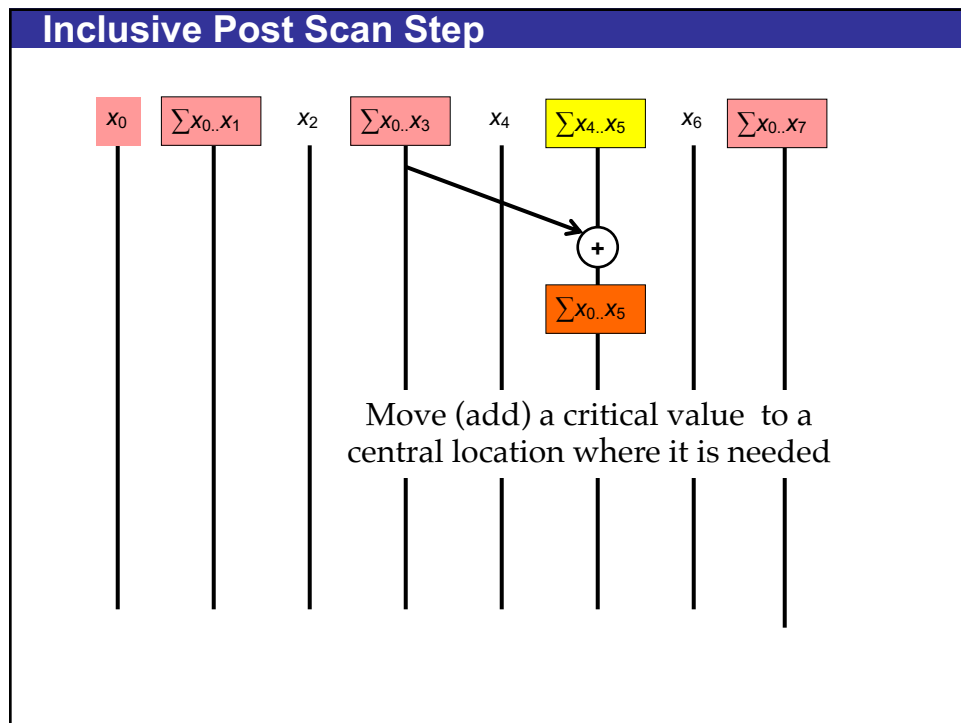
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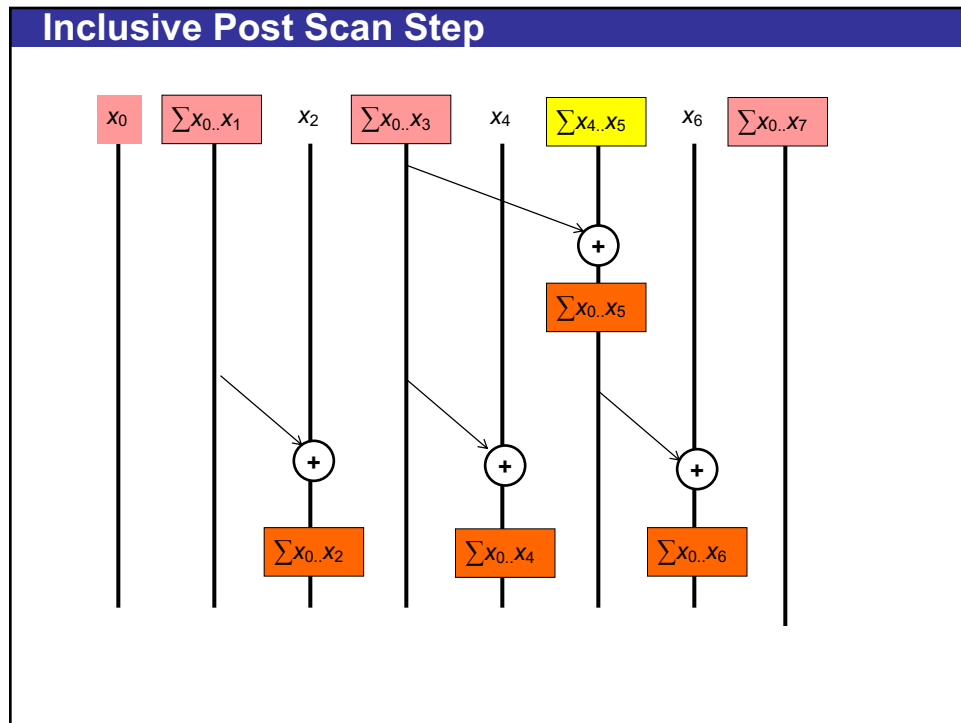
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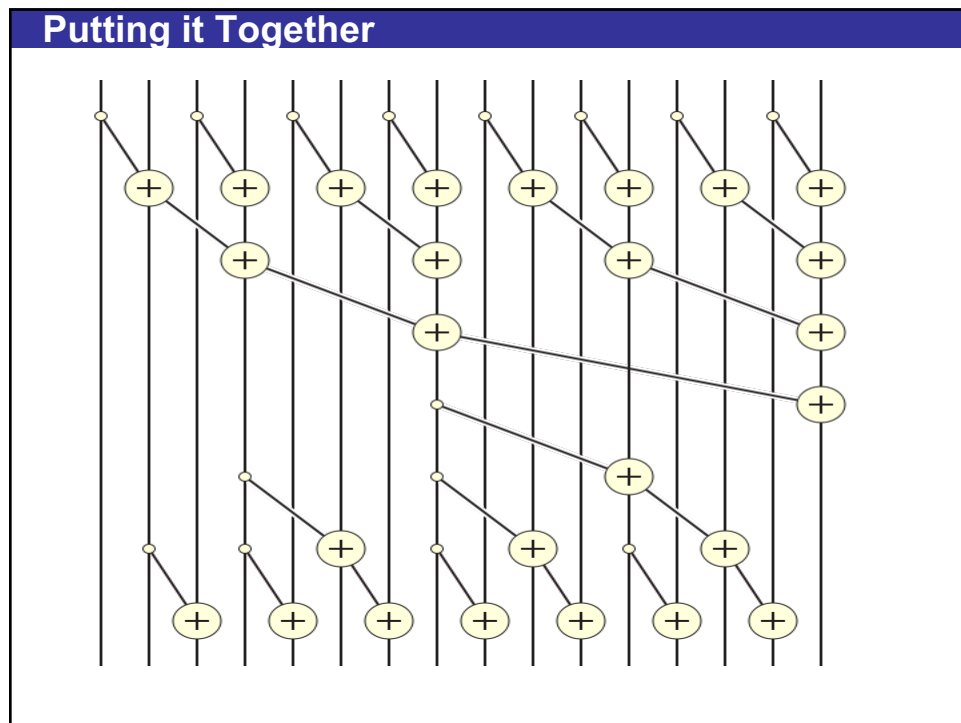
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## Reduction Step Kernel Code

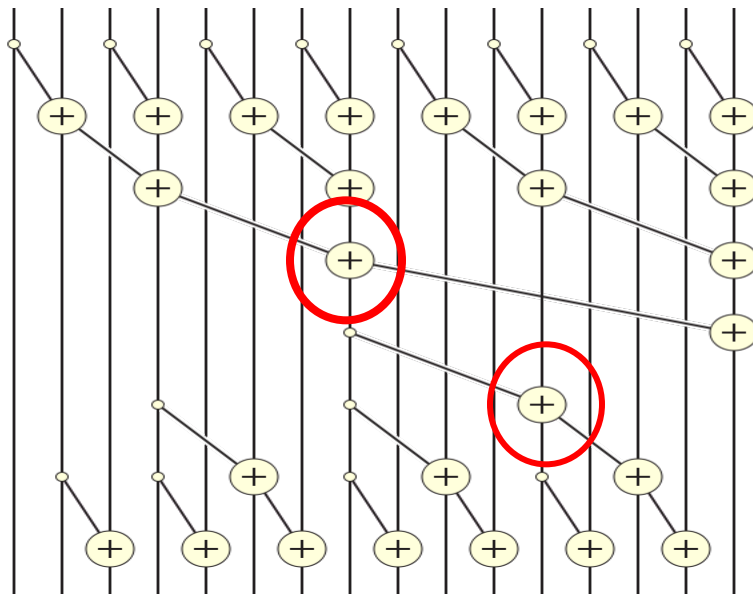
```
// scan_array[BLOCK_SIZE] is in shared memory

int stride = 1;
while (stride < BLOCK_SIZE)
{
    int index = (threadIdx.x+1)*stride*2 - 1;
    if (index < BLOCK_SIZE)
        scan_array[index] += scan_array[index-stride];
    stride = stride*2;

    __syncthreads();
}
```

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## Putting it Together



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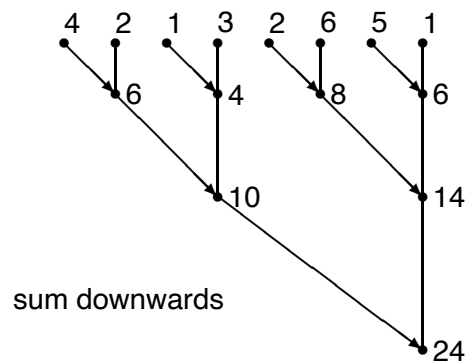


## Post Scan Step

```
int stride = BLOCK_SIZE >> 1;
while(stride > 0)
{
    int index = (threadIdx.x+1)*stride*2 - 1;
    if(index < BLOCK_SIZE) {
        scan_array[index+stride] += scan_array[index];
    }
    stride = stride >> 1;
    __syncthreads();
}
```

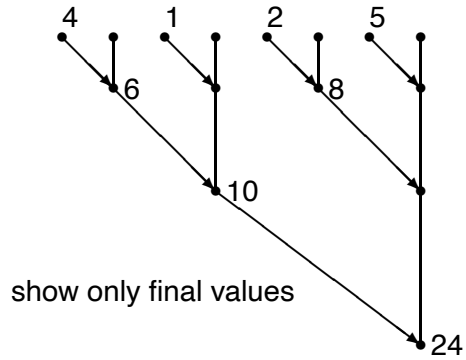
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## Parallel Scan: Another View



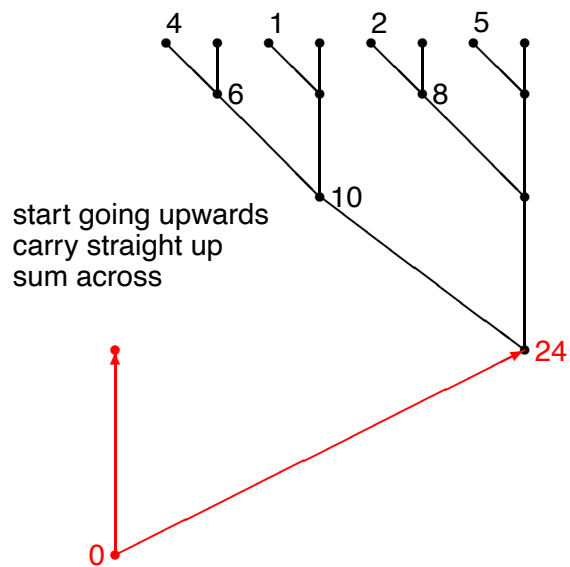
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## Parallel Scan: Another View



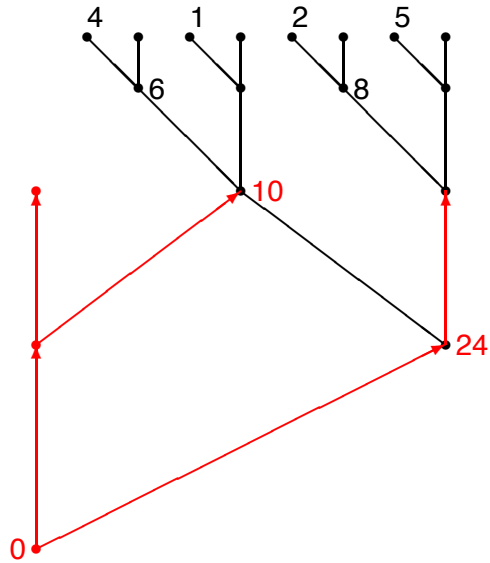
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## Parallel Scan: Another View



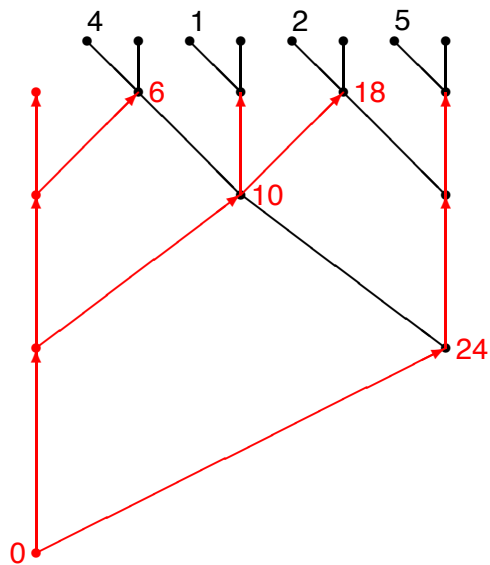
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### Parallel Scan: Another View



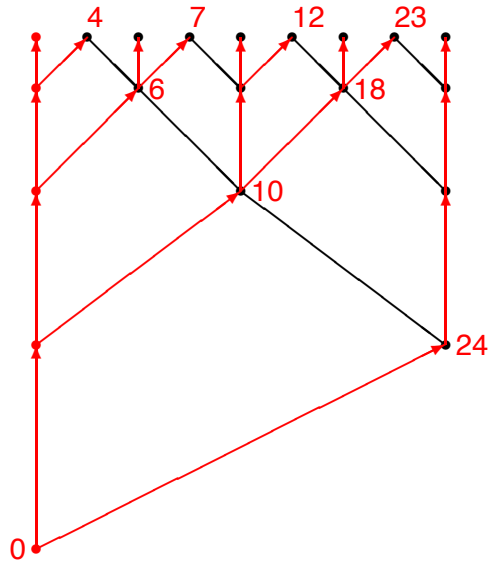
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### Parallel Scan: Another View



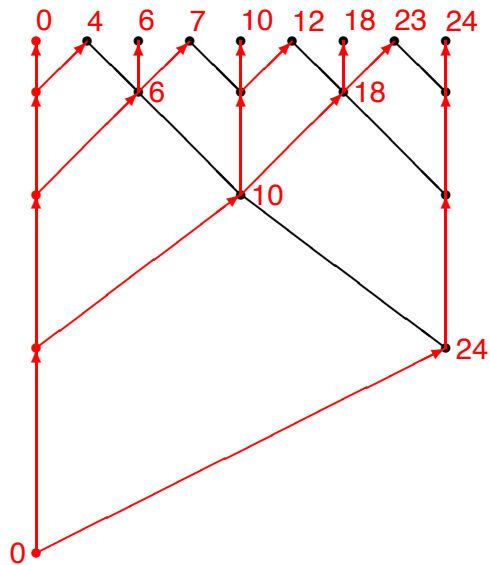
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### Parallel Scan: Another View



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### Parallel Scan: Another View



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### Single-Kernel Parallel Scan: Potential for Deadlock

- However, this needs the sum of all preceding blocks to add to the local scan values
  - replace initial value 0 at the start of the upward sweep
- Problem: blocks are not necessarily processed in order, so could end up in deadlock waiting for results from a block which doesn't get a chance to start.
- Could launch multiple kernels on multiple streams, enforce dependency order using events
  - Dependency encoding might be tricky to get right
  - Switching often to CPU degrades performance
  - Any other solutions?
- **Solution: use atomic increments**

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### Enforcing Block Processing Order

- Declare a global device variable  

```
__device__ int my_block_count = 0;
```
- At the beginning of the kernel code use  

```
__shared__ unsigned int my_blockId;  
if (threadIdx.x==0) {  
    my_blockId = atomicInc( &my_block_count,  
                           UINT_MAX);  
}  
__syncthreads();
```
- This returns the old value of my\_block\_count and increments it, all in one operation. The UINT\_MAX ensures atomicInc always increments the counter.
- This gives us a way of launching blocks in strict order.

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### Block-Ordered Global Scan

- Use a new global counter, `block_sums_completed`, to indicate which blocks have completed their downward sums
- For a single-kernel global scan, the kernel does the following:
  - get in-order block ID; let's call it  $B_i$
  - do downward pass
  - $B_i$  waits until `block_sums_completed` =  $B_i - 1$ . When it does, it shows that the preceding block has computed the sum of the blocks so far on a global variable `sum`
  - get `sum`, increment it with the local partial sum
  - increment `block_sums_completed` to signal to block  $B_{i+1}$  that you are done
  - do upwards pass and store the results

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### (Exclusive) Prefix-Sum (Scan) Definition

**Definition:** The all-prefix-sums operation takes a binary associative operator  $\oplus$ , and an array of  $n$  elements

$$[a_0, a_1, \dots, a_{n-1}],$$

and returns the array

$$[0, a_0, (a_0 \oplus a_1), \dots, (a_0 \oplus a_1 \oplus \dots \oplus a_{n-2})].$$

**Example:** If  $\oplus$  is addition, then the all-prefix-sums operation on the array  $[3 \ 1 \ 7 \ 0 \ 4 \ 1 \ 6 \ 3]$  would return  $[0 \ 3 \ 4 \ 11 \ 11 \ 15 \ 16 \ 22]$ .

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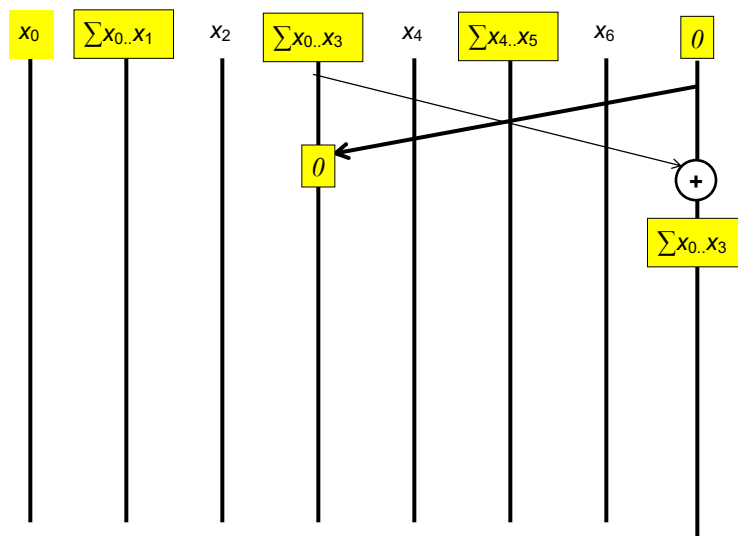
## Why Exclusive Scan

- To find the beginning address of allocated buffers
- Inclusive and Exclusive scans can be easily derived from each other; it is a matter of convenience

	[3 1 7 0 4 1 6 3]
Exclusive	[0 3 4 11 11 15 16 22]
Inclusive	[3 4 11 11 15 16 22 25]

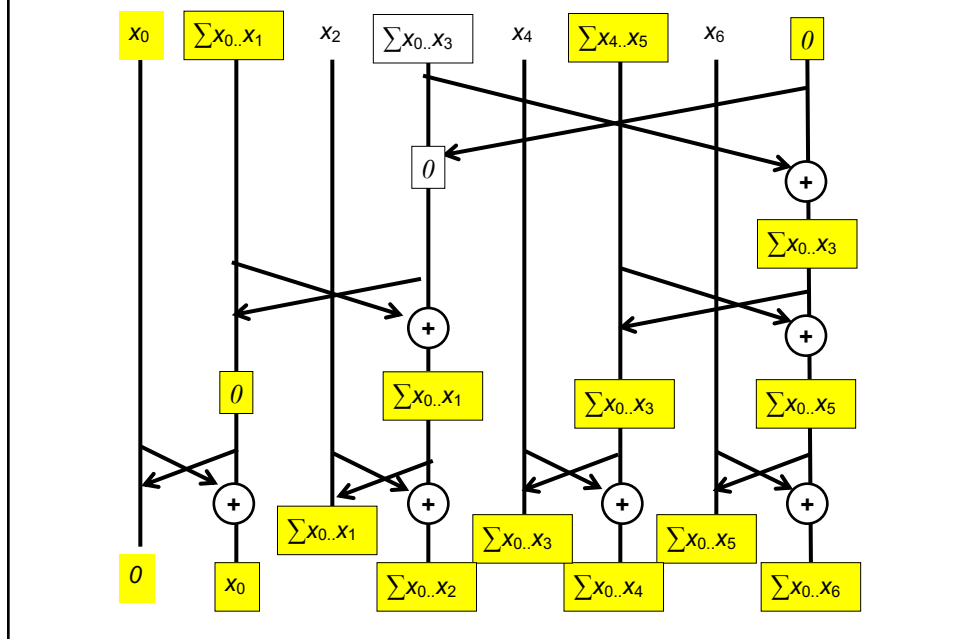
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## Exclusive Post Scan Step



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## Exclusive Post Scan Step



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## Exclusive Scan Example – Reduction Step

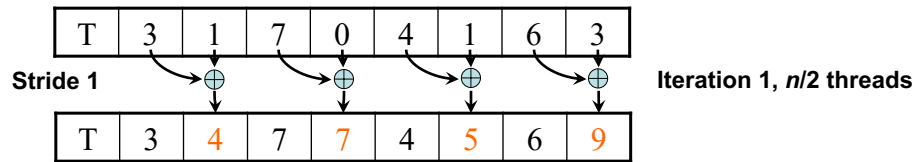
T	3	1	7	0	4	1	6	3
---	---	---	---	---	---	---	---	---

Assume array is already in shared memory

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## Reduction Step (cont.)

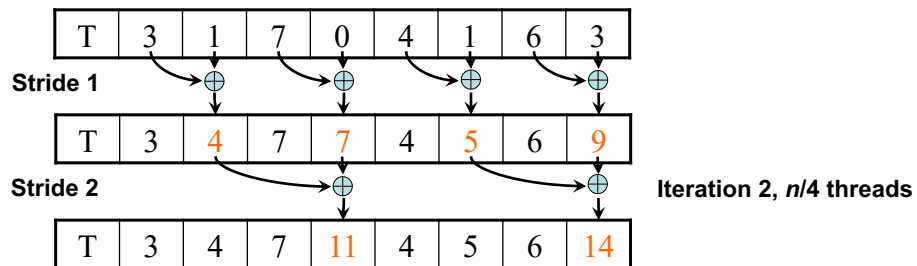


Each  $\oplus$  corresponds to a single thread.

Iterate  $\log(n)$  times. Each thread adds value *stride* elements away to its own value

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## Reduction Step (cont.)

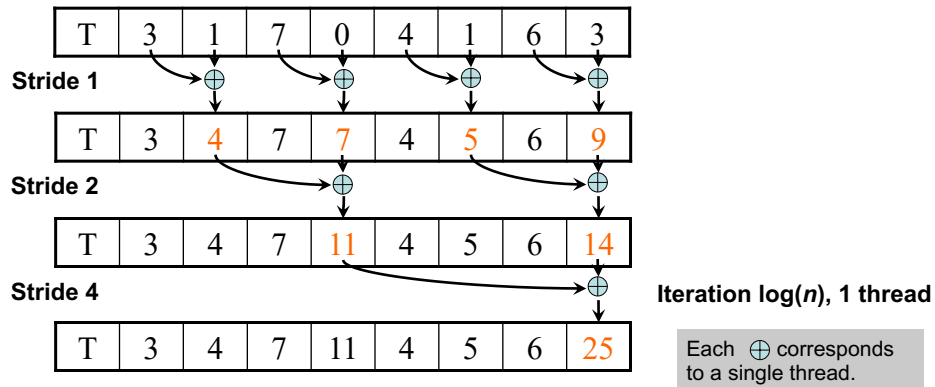


Each  $\oplus$  corresponds to a single thread.

Iterate  $\log(n)$  times. Each thread adds value *stride* elements away to its own value

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## Reduction Step (cont.)



Iterate  $\log(n)$  times. Each thread adds value *stride* elements away to its own value.

Note that this algorithm operates in-place: no need for double buffering

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## Zero the Last Element

T	3	4	7	11	4	5	6	0
---	---	---	---	----	---	---	---	---

We now have an array of partial sums. Since this is an exclusive scan, set the last element to zero. It will propagate back to the first element.

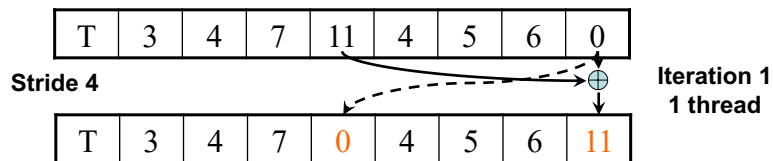
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## Post Scan Step from Partial Sums

T	3	4	7	11	4	5	6	0
---	---	---	---	----	---	---	---	---

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## Post Scan Step from Partial Sums (cont.)

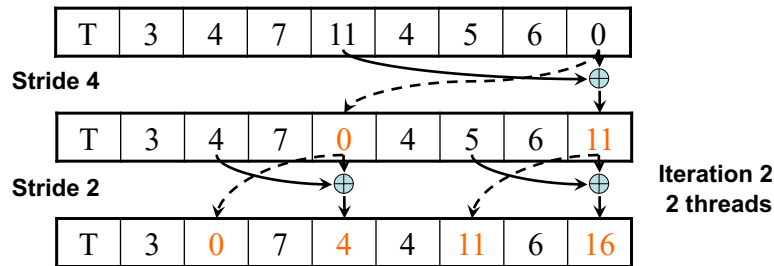


Each  $\oplus$  corresponds to a single thread.

Iterate  $\log(n)$  times. Each thread adds value *stride* elements away to its own value, and sets the value *stride* elements away to its own *previous* value.

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## Post Scan From Partial Sums (cont.)

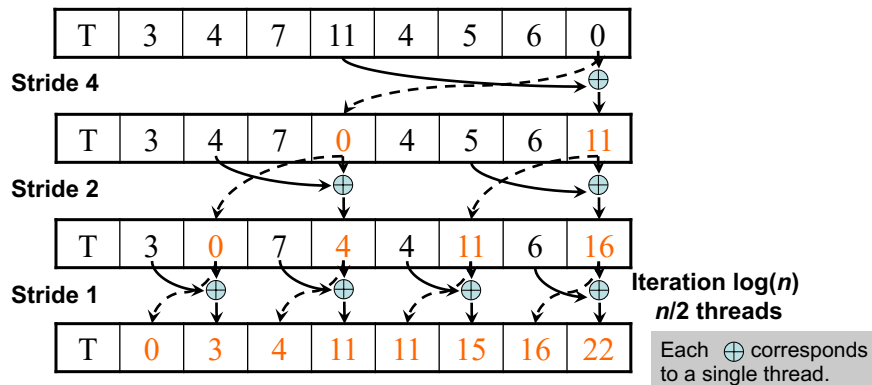


Each  $\oplus$  corresponds to a single thread.

Iterate  $\log(n)$  times. Each thread adds value *stride* elements away to its own value, and sets the value *stride* elements away to its own *previous* value.

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## Post Scan Step From Partial Sums (cont.)



Each  $\oplus$  corresponds to a single thread.

Done! We now have a completed scan that we can write out to device memory.

Total steps:  $2 * \log(n)$ .

Total work:  $2 * (n-1)$  adds =  $O(n)$  **Work Efficient!**

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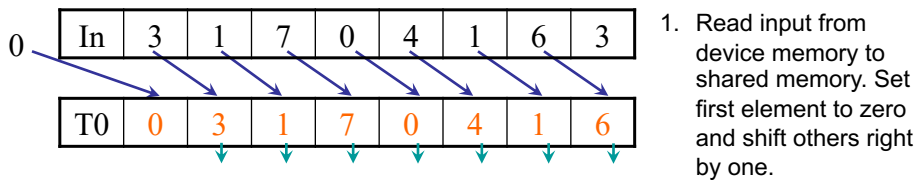
## Work Analysis

- The Parallel Inclusive Scan executes  $2 * \log(n)$  parallel iterations
  - $\log(n)$  in reduction and  $\log(n)$  in post scan
  - The iterations do  $n/2, n/4, \dots, 1, \dots, n/4, n/2$  adds
  - Total adds:  $2 * (n-1) \rightarrow O(n)$  work
- The total number of adds is no more than twice of that done in the efficient sequential algorithm
  - The benefit of parallelism can easily overcome the 2X work

Compare to a work-inefficient parallel scan

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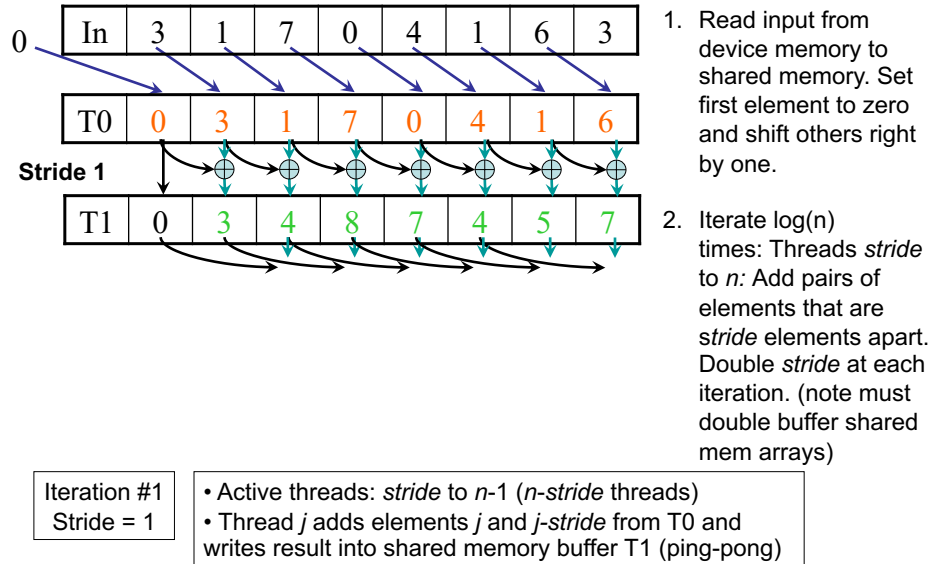
## A Plausible Parallel Scan Algorithm



Each thread reads one value from the input array in device memory into shared memory array T0. Thread 0 writes 0 into shared memory array.

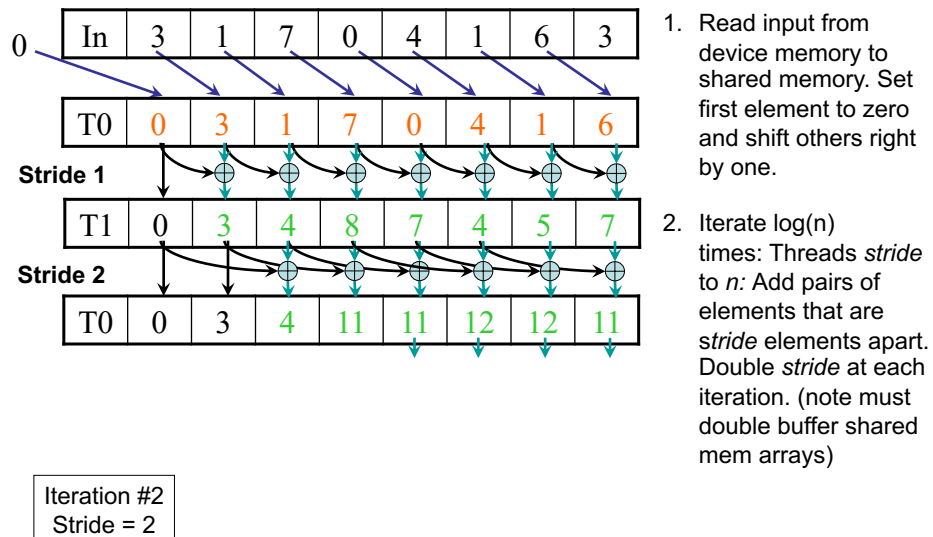
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## A Plausible Parallel Scan Algorithm



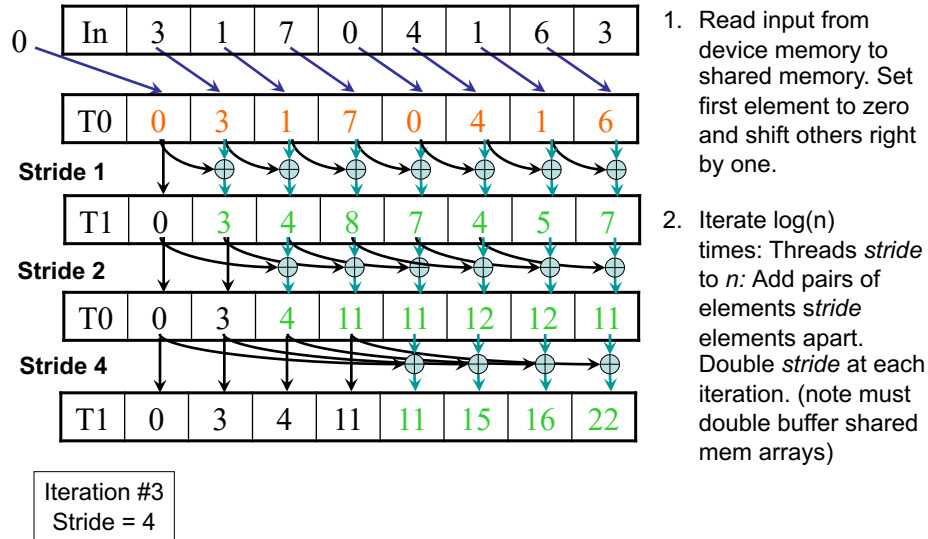
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## A Plausible Parallel Scan Algorithm



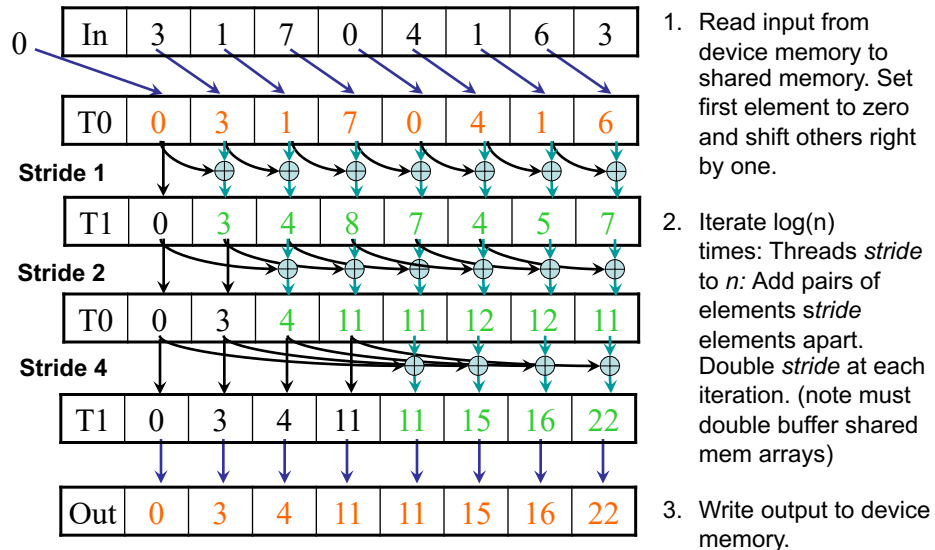
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## A Plausible Parallel Scan Algorithm



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## A Plausible Parallel Scan Algorithm



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### Sample Code for Plausible Parallel Scan Algorithm

```
__global__ void scan(float *d_sum, float *d_data)
{
    extern __shared__ float temp[];
    int tid = threadIdx.x;
    temp[tid] = d_data[tid+blockIdx.x*blockDim.x];
    for (int d=1; d<blockDim.x; d<=1) {
        __syncthreads();
        float temp2 = (tid >= d) ? temp[tid-d] : 0;
        __syncthreads();
        temp[tid] += temp2;
    }
    ...
}
```

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### Work Efficiency Considerations

- The plausible parallel Scan executes  $\log(n)$  parallel iterations
  - The steps do  $(n-1), (n-2), (n-4), \dots, (n - n/2)$  adds
  - Total adds:  $n * \log_2(n) + (n-1) \rightarrow O(n * \log_2(n))$  work
- This scan algorithm is not very work efficient
  - Sequential scan algorithm does  $n$  adds
  - A factor of  $\log_2(n)$  hurts: 20x for  $10^6$  elements!
- A parallel algorithm can be slow when execution resources are saturated due to low work efficiency

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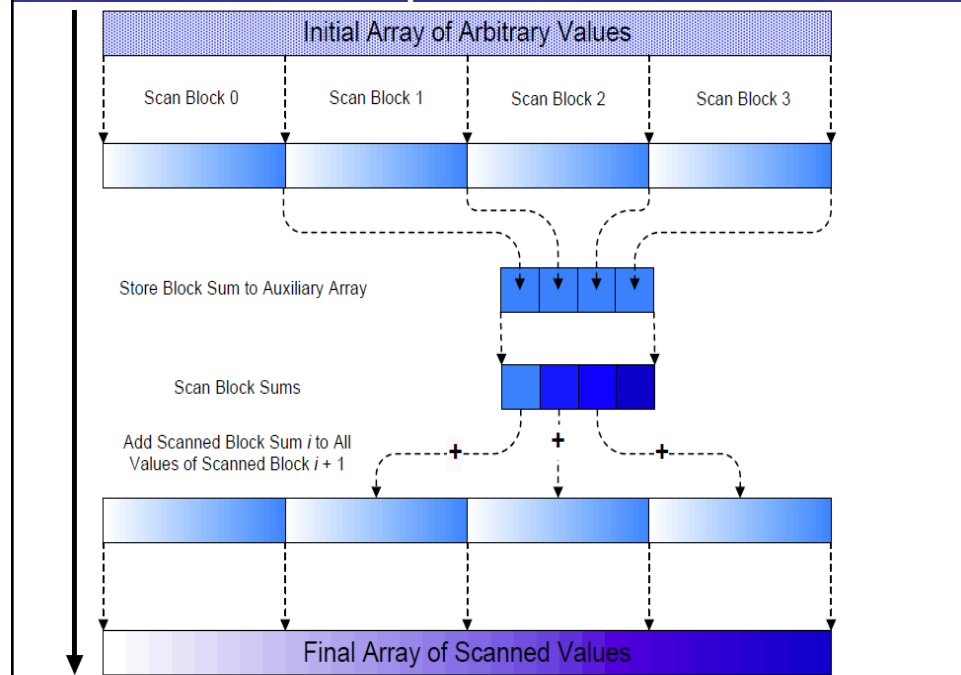


## Working on Arbitrary Length Input

- Build on the scan kernel that handles up to  $2 * \text{blockDim}$  elements
- Have each section of  $2 * \text{blockDim}$  elements assigned to each block
- Have each block write the sum of its section into a Sum array indexed by  $\text{blockIdx.x}$
- Run parallel scan on the Sum array
  - May need to break down Sum into multiple sections if it is too big for a block
- Add the scanned Sum array values to the elements of corresponding sections

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## Overall Flow of Complete Scan



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