Chapter 3: Parallel Algorithmic Structures I

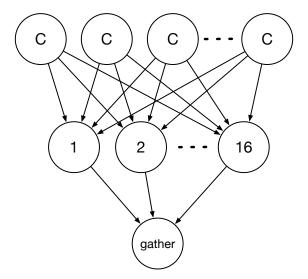
Elements of Parallel Computing

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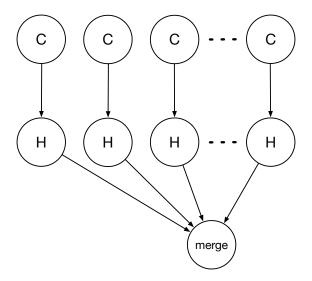
Histogram Example

- Large collection of short messages from social network
- Want to produce histogram showing distribution of complexity of messages
- Assumptions:
 - complexity estimated as fraction of letters in alphabet that occur in a message,
 - histogram contains 16 bins covering range of complexities from 0 to 1

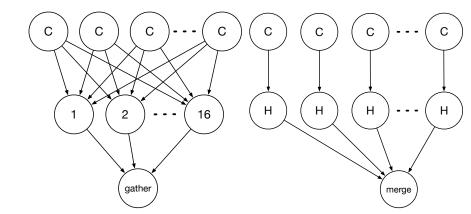
Histogram Decomposition #1



Histogram Decomposition #2



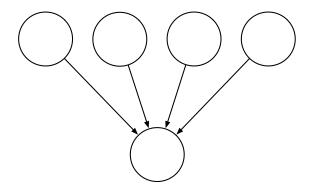
Histogram Decompositions



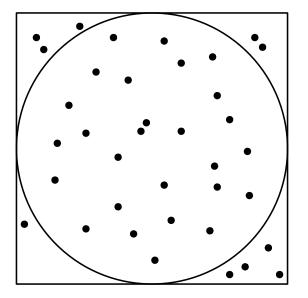
Guidelines for Parallel Algorithm Design

- 1. Postpone consideration of the details of the computational platform until after the decomposition phase.
- 2. Create many independent tasks, whose number increases with the problem size.

Embarrassingly Parallel

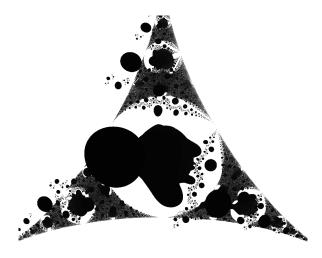


Monte Carlo estimation of π



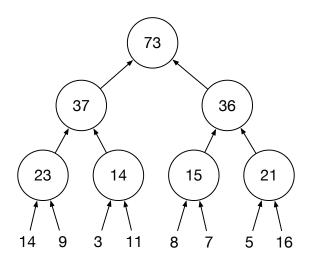
Generalized Fractal

- Recurrence reliation: $z \leftarrow z^{\alpha} + c$
 - ightharpoonup α real and z complex
 - initial value (0,0) for $\alpha > 0$ and (1,1) for $\alpha < 0$
 - Coordinates of image are real and imaginary parts of c
- Number of iterations to diverge mapped to a colour
 - grayscale: white don't diverge, darker the gray the earlier the divergence

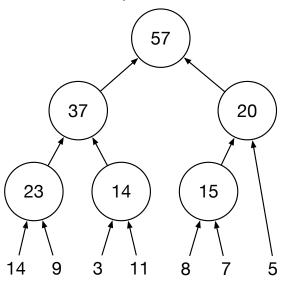


Find the workload imbalance

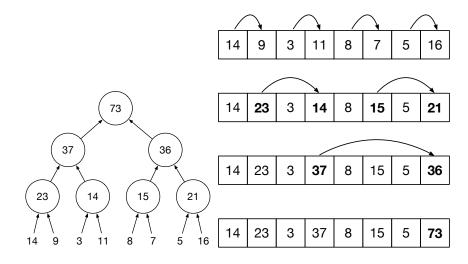
Reduction



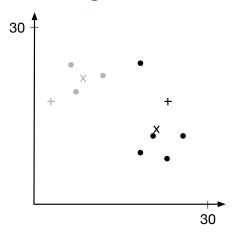
Power of 2 Not Required



Two Views of Reduction



K-means Clustering



+: initial guess, \times : centroid of points assigned to each cluster

K-means Clustering Algorithm

Input: array of *n*-dimensional vectors, and the number of clusters (*k*)

Output: array *closest* containing assignment of each vector to one of the clusters

```
1 Assign initial guesses of the k cluster centers cluster[k]
   Initialize clusterNew[k] and clusterSize[k] arrays to zero
   while not converged do
        foreach vector i do
             find cluster center i with smallest Euclidean distance
 5
             to vector i
             closest[i] \leftarrow i
 6
             clusterNew[i] \leftarrow clusterNew[i] + vector[i]
             clusterSize[i] \leftarrow clusterSize[i] + 1
 8
        end
 9
        foreach cluster center i do
10
             cluster[i] \leftarrow clusterNew[i]/clusterSize[i]
11
             clusterNew[i] \leftarrow 0
12
             clusterSize[i] \leftarrow 0
13
        end
14
15 end
```

K-means Example

Iteration over vectors:

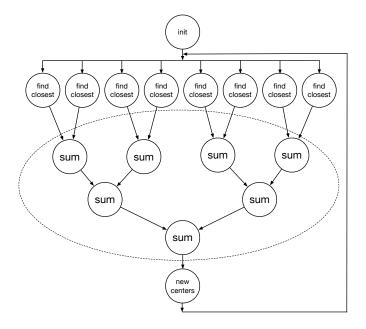
clusterNew array:

$$\begin{aligned} & [[6.2,23.9],0] + [[7.1,19.3],0] + [[11.7,22.1],0] + \\ & [0,[18.1,24.2]] + [0,[18.1,8.9]] + [0,[20.3,11.8]] + \\ & [0,[22.7,7.9]] + [0,[25.4,11.8]] \end{aligned}$$

clusterSize array:

$$[1,0]+[1,0]+[1,0]+[0,1]+[0,1]+[0,1]+[0,1]+[0,1]$$

closest array: [0, 0, 0, 1, 1, 1, 1, 1]



Scan

Scan: binary associative operator to an *ordered* collection of values, produces new collection

- inclusive: element i contains the result of the operator applied to the first i elements of the original collection.
- exclusive: element i contains the result of the operator applied to the first i-1 elements of the original collection.

Prefix Sum

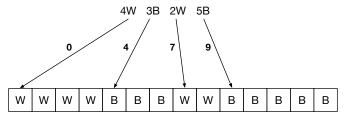
Prefix sum: scan with sum operator

- ► Inclusive prefix sum of [2, 16, 9, 7]: [2, 18, 27, 34].
- ► Exclusive prefix sum of [2, 16, 9, 7]: [0, 2, 18, 27].

Good Use of Scan: Parallel Write to Array

Run-length encoding: WWWWBBBWWBBBBB represented as 4W3B2W5B

Parallel run-length decoding with 4 tasks: starting point for each write of each task from exclusive prefix sum of [4, 3, 2, 5]: [0, 4, 7, 9]



How to Parallelize Scan?

```
sum \leftarrow 0

for i \leftarrow 0 to n-1 do

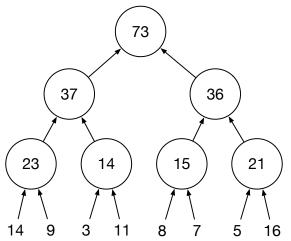
sum \leftarrow sum + a[i]

scan[i] \leftarrow sum

end
```

Blelloch Scan (1)

Start with pass up reduction tree:



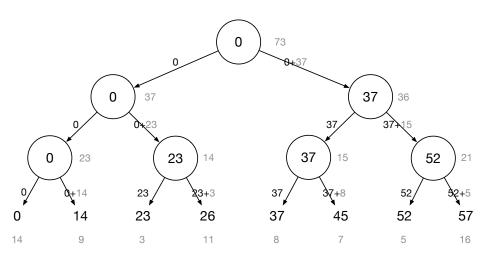
Blelloch Scan (2)

Back down tree to compute scan:

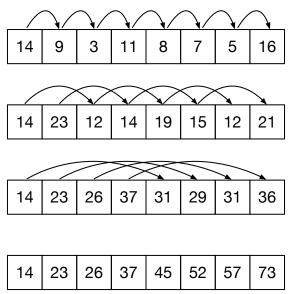
```
scan[root] \leftarrow 0

scan[left[v]] \leftarrow scan[v]

scan[right[v]] \leftarrow scan[v] + reduce[left[v]]
```



Hillis and Steele Scan



Task Graph for Hillis and Steele

