All-Pairs-Shortest-Paths in Spark

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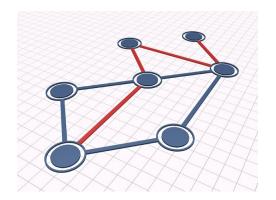
Stanford University

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Problem

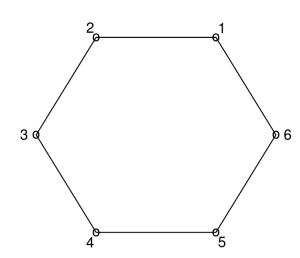
- Weighted graph G = (V, E) with n vertices
- Compute $n \times n$ matrix of distances S where

 S_{ij} = weight of shortest path from i to j

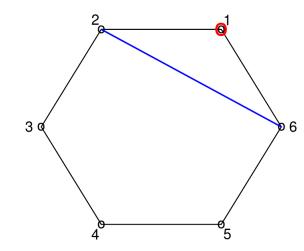


Initial input

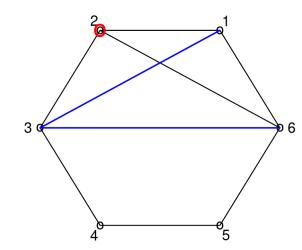
$$\begin{pmatrix} 0 & 1 & & & 1 \\ 1 & 0 & 1 & & & \\ & 1 & 0 & 1 & & \\ & & 1 & 0 & 1 \\ & & & 1 & 0 & 1 \\ 1 & & & & 1 & 0 \end{pmatrix}$$

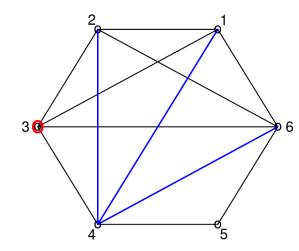


$$\begin{pmatrix} 0 & 1 & & & 1 \\ 1 & 0 & 1 & & & \mathbf{2} \\ & 1 & 0 & 1 & & & \\ & & 1 & 0 & 1 & & \\ & & & 1 & 0 & 1 \\ 1 & \mathbf{2} & & & 1 & 0 \end{pmatrix}$$

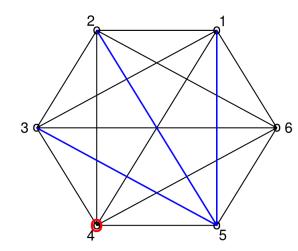


$$\begin{pmatrix} 0 & 1 & \mathbf{2} & & 1 \\ 1 & 0 & 1 & & 2 \\ \mathbf{2} & 1 & 0 & 1 & \mathbf{3} \\ & & 1 & 0 & 1 \\ & & & 1 & 0 & 1 \\ 1 & 2 & \mathbf{3} & & 1 & 0 \end{pmatrix}$$

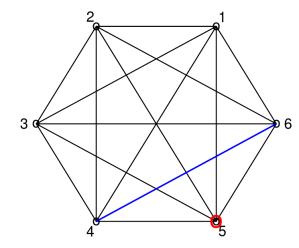




$$\begin{pmatrix} 0 & 1 & 2 & 3 & \mathbf{4} & 1 \\ 1 & 0 & 1 & 2 & \mathbf{3} & 2 \\ 2 & 1 & 0 & 1 & \mathbf{2} & 3 \\ 3 & 2 & 1 & 0 & 1 & \mathbf{4} \\ \mathbf{4} & \mathbf{3} & \mathbf{2} & 1 & 0 & 1 \\ 1 & 2 & 3 & \mathbf{4} & 1 & 0 \end{pmatrix}$$

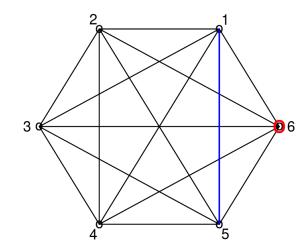


$$\begin{pmatrix} 0 & 1 & 2 & 3 & 4 & 1 \\ 1 & 0 & 1 & 2 & 3 & 2 \\ 2 & 1 & 0 & 1 & 2 & 3 \\ 3 & 2 & 1 & 0 & 1 & 2 \\ 4 & 3 & 2 & 1 & 0 & 1 \\ 1 & 2 & 3 & 2 & 1 & 0 \end{pmatrix}$$



Iteration 6, (terminate)

$$\begin{pmatrix} 0 & 1 & 2 & 3 & \mathbf{2} & 1 \\ 1 & 0 & 1 & 2 & 3 & 2 \\ 2 & 1 & 0 & 1 & 2 & 3 \\ 3 & 2 & 1 & 0 & 1 & 2 \\ \mathbf{2} & 3 & 2 & 1 & 0 & 1 \\ 1 & 2 & 3 & 2 & 1 & 0 \end{pmatrix}$$

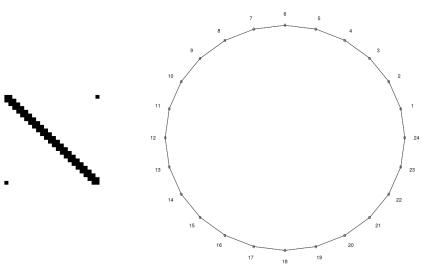


- Cost: $O(n^3)$ operations (single-core)
- Takes *n sequential* iterations

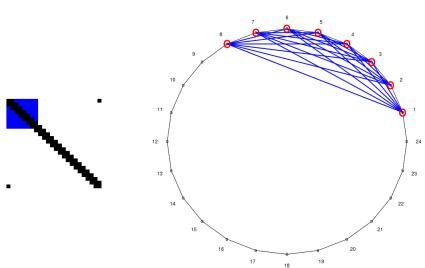
Problems with Floyd-Warshall

- FW updates using 1 vertex at a time
- Result: *n* iterations
- High # iterations = latency in distributed setting
- Solomonik et al. (2013) show how to "block" FW iterates
- We modify their block-based approach for Spark

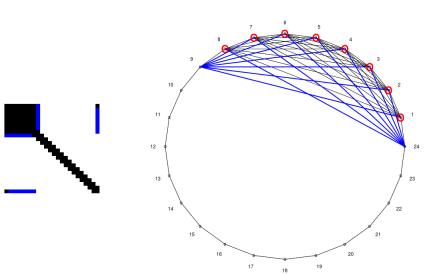
Initial input: n = 24, block size = 8



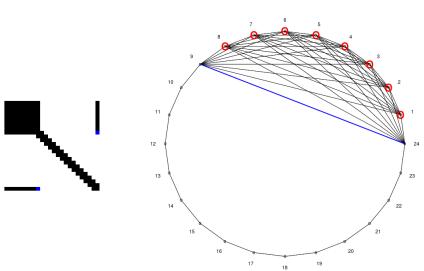
Iteration 1A: Update all paths within block 1 (with FW)



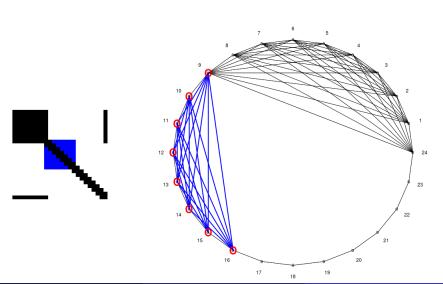
Iteration 1B: Update all paths to/from block 1



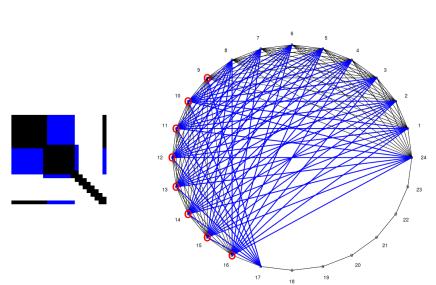
Iteration 1C: Update all paths



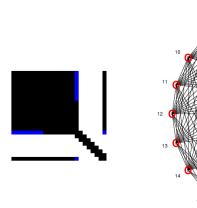
Iteration 2A: Update all paths within block 2 (with FW)

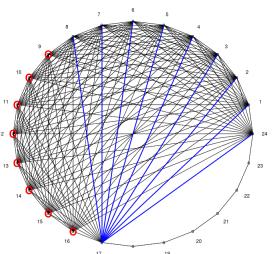


Iteration 2B: Update all paths to/from block 2

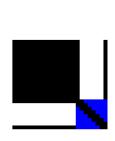


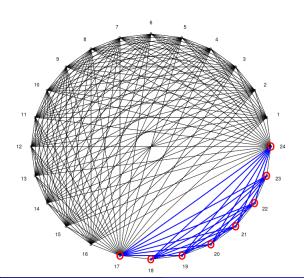
Iteration 2C: Update all paths



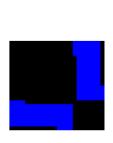


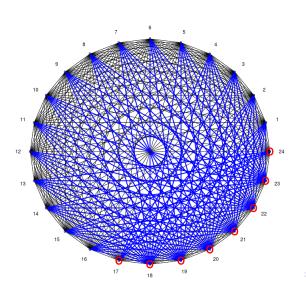
Iteration 3A: Update all paths within block 3 (with FW)





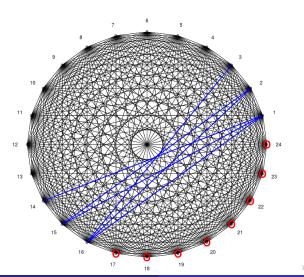
Iteration 3B: Update all paths to/from block 3





Iteration 3C: Update all paths, (terminate)





Block APSP: Single-core

- Block size b, n/b iterations
- A-step (all paths within block): $O(b^3)$
- B-step (all paths to/from block): $O(nb^2)$
- C-step (all paths): $O(n^2b)$
- Iteration: $O(n^2b + nb^2 + b^3)$
- Total: $O(\frac{n}{b}(n^2b + nb^2 + b^3)) = O(n^3 + n^2b + nb^2)$
- ullet The case b=1 is almost the same as Floyd-Warshall

Distributing Block APSP

Problem setup

- Input format: Given by dense adjacency matrix, stored as BlockMatrix S with block size b
- Output format: same
- Let S[i,j] denote the i,jth block of S
- Each worker holds k contiguous blocks
- Set block size b so that $k \frac{n^2}{b^2}$ fits in memory

Scaling

- Number of vertices $n \to \infty$
- Number of workers $p = C_w n^2$
- Block size b, blocks per worker k are constant

Distributing Block APSP

For
$$i = 1, \ldots, n/b$$

- A-step: (update all paths within block)
 - filter + collect diagonal submatrix S[i,i]
 - Driver computes X = Floyd-Warshall(S) locally
 - Communication cost: $O(b^2)$
- B-step: (update all paths to/from block)
 - filter + mapValues: update b rows and columns using X
 - One-to-all communication (broadcast X)
 - Communication cost: $O(b^2 \log(p))$
- C-step: (update all paths)
 - flatMap to duplicate b rows/columns and join to send updated rows/columns to every block
 - map to update each block with updated rows/columns
 - All-to-all communication (join)
 - Communication cost: $O(kb^2p \log p)$



Distributing Block APSP

Overall cost:

- Total computational cost is $O(n^3)$, divided evenly among workers
- Total communication cost: $O(nb(kp+1)\log p)$
- Total latency/synchronization cost: $O(\frac{n}{b}\log(p))$

Optimal performance when b is as large as possible, e.g.

$$b = \sqrt{\frac{\mathsf{RAM}}{3}}$$

Hence should be much better than Floyd-Warshall (b=1)

Results