

Assignment #4

Group 4 presents solutions at **2/12**

Group 2 prepares questions/discussion

Sparse matrix – vector multiplication

- Implement the unstructured, sparse matrix – vector multiplication from *Bolz et al. Sparse Matrix Solvers on the GPU: Conjugate Gradients and Multigrid*.
 - You may assume there is exactly two entries per row and no diagonal elements - but do not optimize away R^x .
 - Validate your results

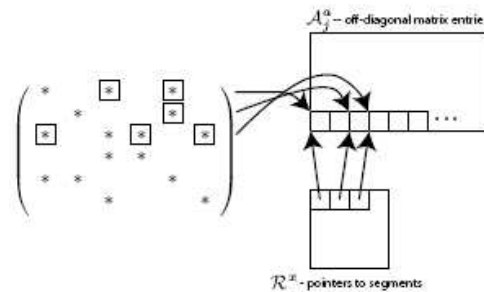


Figure 1: Off-diagonal elements of each row are compacted into segments which are tightly packed into A_j^α . A pointer to the beginning of each segment is stored in R^x .

$$\begin{aligned}
 j &= R^x[i] \\
 Y^x[i] &= A_i^\alpha[i] * X^x[i] + \sum_{c=0}^{h_i-1} A_j^\alpha[j+c] * X^x[C^\alpha[j+c]].
 \end{aligned}$$

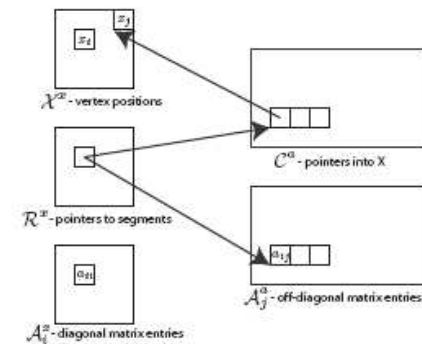


Figure 2: When the fragment program executes on the pixel corresponding to row i , the window position is used as a texture coordinate to fetch x_i in X^x and a_{ii} in A_i^α . The window position also identifies the segment pointer in R^x , which points to the location of non-zero elements a_{ij} in A_j^α corresponding to row i . Finally, using the segment pointer from R^x we can access entries in C^α which reveal the addresses of x_j in X^x corresponding to non-zero a_{ij} .



Optional, but please try make it

- Implement the structured, sparse matrix – vector multiplication from *Krüger and Westermann. Linear algebra Operators for GPU implementation of Numerical Algorithms*.
 - You may assume there is exactly two bands
 - Validate your results
- Compare the speed of the two methods of matrix-vector multiplication
 - In our algorithm design, does it make sense to aim for either structured or unstructured sparse matrices specifically?