Blue Waters Petascale Semester Curriculum v1.0

Unit 4: OpenMP

Lesson 8: Markov chains, Matrix multiply

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for the Shodor Education Foundation, Inc.

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Markov Chains

SERIAL AND PARALLEL MATRIX MULTIPLICATION

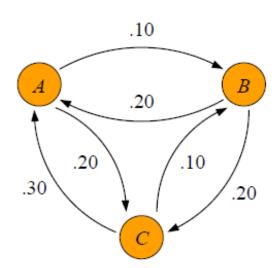
Learning Objectives

- ▶ The student should understand
 - ▶ The concept of a Markov Chain
 - ► The transition matrix
 - State changes using the transition matrix and a state vector (matrix-vector multiplication)
 - The steady state (iterated matrix-matrix multiplication)
 - How simple it can be to convert a serial program into a parallel program using OpenMP

Markov Chain

- ► A mathematical representation of a problem involving state transitions
- ▶ The transitions are based on a fixed probability
 - When the system is in one state the probability of moving to another state is fixed
- Useful to visualize a Markov chain with a state space diagram
 - Each state is represented as a node (vertex or circle)
 - Probabilities of transitioning from one state to another state
 - ► Weighted edges between the nodes

- ▶ Three vats of water, A, B, and C
 - ▶ 10% of the water in A is transferred into B and 20% in A is transferred into to C in one hour
 - ▶ 20% of the water in B is transferred into A and 20% from B is transferred into C in one hour
 - ▶ 30% of the water in C is transferred into A and 10% from C is transferred into B in one hour
- ▶ The state-space diagram



- The question to be answered is, how much water is in each vat after a certain number of hours
- The state space diagram can be represented as a two-dimensional matrix, called a transition matrix
 - ▶ The columns represent the transition amounts, or the probabilities
 - ▶ The columns must sum to one

$$T = \begin{bmatrix} 0.7 & 0.2 & 0.3 \\ 0.1 & 0.6 & 0.1 \\ 0.2 & 0.2 & 0.6 \end{bmatrix}$$

▶ T[r][c] represents the percentage of water that is taken from vat c and transferred to vat r

- A vector is used to represent the initial amount of water in each vat
- For example, let vat A initially contain 5 gallons of water, vat B contain 10 gallons, and vat C contain 15 gallons

$$V_0 = \begin{bmatrix} 5 \\ 10 \\ 15 \end{bmatrix}$$

The amount of water in each vat is determined by multiplying the transition matrix by the vector, or

$$V_1 = TV_0$$

$$V_1 = \text{TV}_0 = \begin{bmatrix} 0.7 & 0.2 & 0.3 \\ 0.1 & 0.6 & 0.1 \\ 0.2 & 0.2 & 0.6 \end{bmatrix} \begin{bmatrix} 5 \\ 10 \\ 15 \end{bmatrix} = \begin{bmatrix} 10 \\ 8 \\ 12 \end{bmatrix}$$

- ▶ The previous equation states that after one hour vat A will contain 10 gallons, vat B will contain 8 gallons, and vat C will contain 12 gallons
- ► The amount of water in each vat after the next hour is found in a similar manner, using these new quantities as the initial conditions

$$V_2 = \text{TV}_1 = T[TV_0] = \begin{bmatrix} 0.7 & 0.2 & 0.3 \\ 0.1 & 0.6 & 0.1 \\ 0.2 & 0.2 & 0.6 \end{bmatrix} \begin{bmatrix} 10 \\ 8 \\ 12 \end{bmatrix} = \begin{bmatrix} 12.2 \\ 7 \\ 10.8 \end{bmatrix}$$

Likewise, the amount of water in each vat after N hours is

$$V_n = T \dots [TV_0] = T \dots TV_0 = T^N V_0$$

► The previous equation demonstrates it is not necessary to computer each state to determine the state after N hours, but rather you need to multiply the transition matrix by itself N times

Matrix Multiplication

- When multiplying two matrices together, the number of columns of the matrix on the left must be the same as the number of rows of the matrix on the right
- ► The final result of multiplying two matrices together is a third matrix, and the process can be described by the equation,

$$C = AB$$

- Where A, B, and C are matrices
- ► Each element in the product matrix is formed by multiplying all the elements of a row with the corresponding elements of a column
- For example,

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}, B = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$

Matrix Multiplication

Then

$$C = \begin{bmatrix} 1(5) + 2(7) & 1(6) + 2(8) \\ 3(5) + 4(7) & 3(6) + 4(8) \end{bmatrix} = \begin{bmatrix} 19 & 22 \\ 43 & 50 \end{bmatrix}$$

Which can be written mathematically as

$$c_{i,j} = \sum_{k=0}^{n-1} a_{i,k} b_{k,j}$$

- ► Each element in the C matrix requires *n* multiplications
- Matrix multiplication is an n^3 operation, which is prohibitively expensive for large matrices

Matrix Multiplication

- ▶ The code to perform this computation is straightforward
 - For all rows and all columns, compute each element in the resulting matrix
 - ▶ This code assumes the matrix C is initialized with zeros

```
for (int r = 0; r < n; r++)

for (int c = 0; c < n; c++)

for (int k = 0; k < n; k++)

C[r][c] += A[r][k] * B[k][c];
```

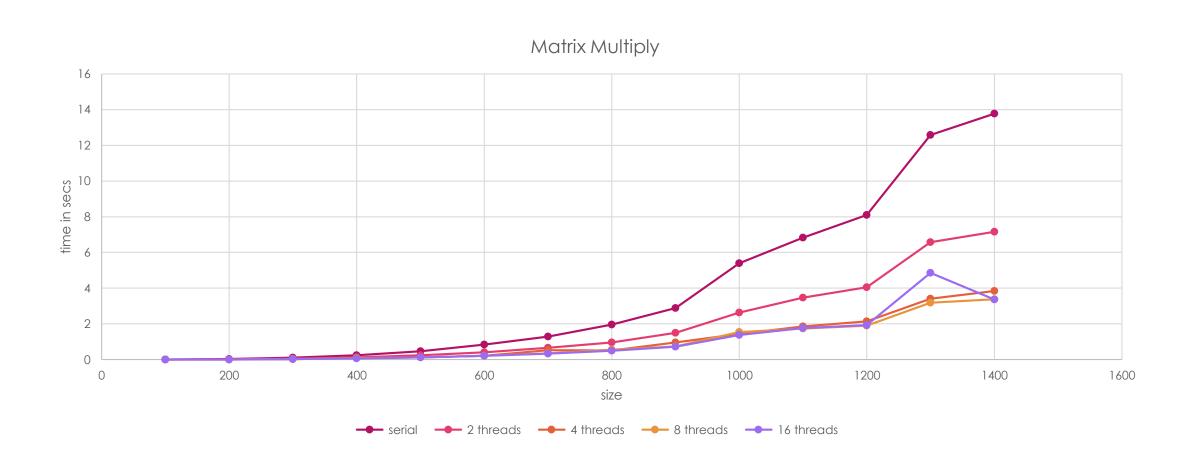
Parallel Matrix Multiplication

- Matrix multiplication is an example of an embarrassingly parallel problem
 - All computations are independent and can therefore be performed simultaneously
- The power of OpenMP is to simply convert serial code for an embarrassingly parallel problem into a parallel solution by adding a single directive to the serial code

Parallel Matrix Multiplication

- ► The pragma tells the compiler to break up the A matrix into a number of blocks containing a similar number of rows, and have independent threads perform the computation on each block of rows simultaneously
 - ► Each thread performs fewer computations compared to the serial case, therefore improving the performance of the matrix multiply code
 - There is some overhead of performing the computation in parallel
 - ▶ The speedup is less than one over the number of threads
 - ▶ The speedup becomes more significant as the size of the matrix increases
 - ▶ The number of hardware threads also limits any performance improvement
- The following plot shows the time to perform matrix multiplication on various sized matrices using different numbers of threads

Parallel Matrix Multiplication Efficiency



Conclusion

- Matrix multiplication is the fundamental processing step when a problem can be mathematically modeled with a Markov Chain
- Matrix multiplication is an embarrassingly parallel computational problem that can be easily made parallel with OpenMP
- Dramatic improvements in computation time can be observed with a parallel version of matrix multiplication, especially as the size of the matrices get large