

Watts-Stogatz Experiments on CArray and KokkosCArrays

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1 Introduction

In this report I detail the methods and results of the floyd test. In this test I generate 1000 node watts-Stogatz graph with different parameters, run floyd warshall over this graph, and report the average shortest path distance over the graph.

1.1 But what is a Watts-Stogatz Graph

A Watts-Stogatz graph is a random graph model introduced in the paper Collective dynamics of small-world networks by Watts and Stogatz. This graph type has many interesting properties such as the clustering coefficient (how many 3 vertex triples are connected triangles) and for this experiment the average shortest distance has been well studied.

A watts-Stogatz graph, forever more in this report a WS graph, is created by the following process with parameters n , k , and p . 1. Make an empty n vertex matrix 2. For each node i connect i with a directed edge to its nearest $2k$ neighbors. This forms a ring lattice. 3. For each edge rewire it with probability p to a random vertex 3b. For practical reasons we dis allow self loops and repeated edges

The elegance of this graph is showing that in a world where most people or entities have a lot of local connections even a very low number of “far” edges to anywhere in the network can dramatically decrease the average shortest path in the network. This sort of graph is common in network. The behavior that many networks exhibit of having a surprisingly small average shortest path is a phenomenon that this relatively simple model hopes to capture.

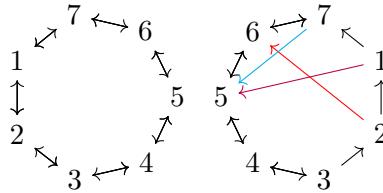


Figure 1: On the left, ring lattice with 7 elements where each node is connected to its nearest 1 neighbor. On the right we replace three edges with random edges. In particular we replaced $(1,2)$ with $(1,5)$, $(2,3)$ with $(2,6)$ and $(7,1)$ with $(7,5)$. This is a simple showcase of the Watts-Stogatz random graph process.

2 Expected distance calculation

When there is no rewiring, that is each node is connected to the k nearest nodes on each side, calculating the average distance is easy. First we observe that we only ever have to go half way around the ring lattice to get to any other node so we will calculate the average distance of the $N/2$ nodes clockwise from us. The first K are only one step away, the next K are 2 steps, and so on. The last bunch will be $\text{ceiling}(\frac{N}{2k})$ steps away. So we get that the average distance is

$$\frac{2}{N} \sum_{i=1}^{\text{ceiling}(\frac{N}{2k})} k \approx \frac{2}{N} k \frac{N(N+1)}{8k^2} \quad (1)$$

$$= \frac{N+1}{4k} \quad (2)$$

$$\approx \frac{N}{4k} \quad (3)$$

On the other extreme if $p=1$ then this graph looks like an erdos renyi graph where each node has a fixed out degree. The expected distance in an Erdos-Renyi graph model is logarithmic in terms of N so we should expect something similar.

3 The experiments

The primary expect is to test MATAR CArrays and kokkos enabled CArrays for speed with varying problem size. In this test case we initialize a WS graph by assigning the local edges, flipping a weighted coin for every edge corresponding to the rewire probability in the times we do rewire we change this edge from a local edge to a random edge. There were no safeguards put in place to make sure double edges are prevented and in theory this makes walks slightly longer. After initializing this graph, we run the floyd warshall algorithm to compute the shortest path length for each pair of vertices. This is an $O(n^3)$ algorithm.

```
G = // adjacency matrix of graph
res = // Graph with 0 on diagonal, values of G on off diagonal and infinity elsewhere
for k in range(0,n):
    for i in range(0,n):
        for j in range(0,n):
            if (res(i,j) > res(i,k) + res(k,j)):
                res(i,j) = res(i,k) + res(k,k)
return res
```

4 Time results

For the time section the rewire probability is 0.00 and we connected to the 12 nearest nodes(i.e the 6 nearest on each side of the node).

Note that since this algorithm is $O(n^3)$ a double of the node size should correspond to an 8x the amount of work being done so the parallel version of the code is a better than it first appears.

Node Size vs Time

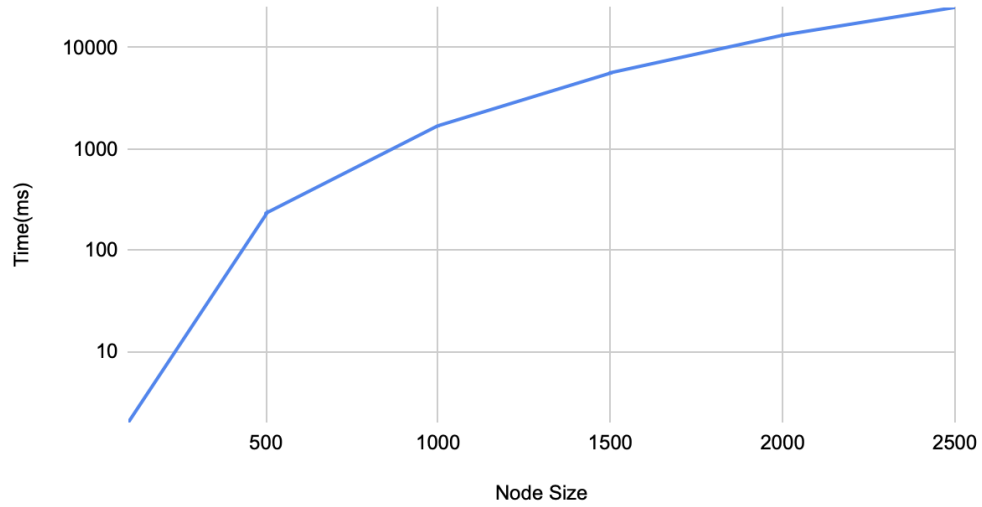


Figure 2: Time taken versus running this process on the serial CArrays

Node Size vs time(ms)

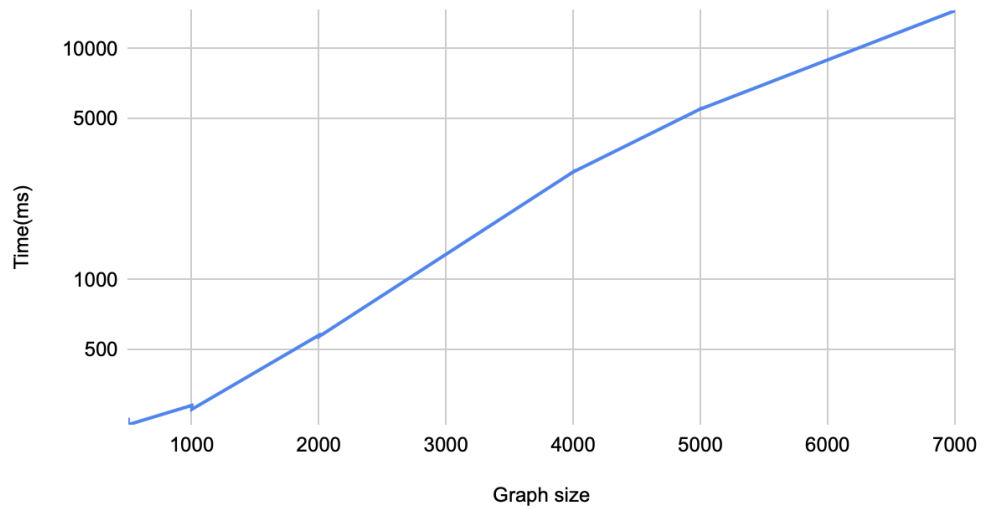


Figure 3: Time taken using the CArrayKokkos running in parallel on the gpu

Comparative runtimes

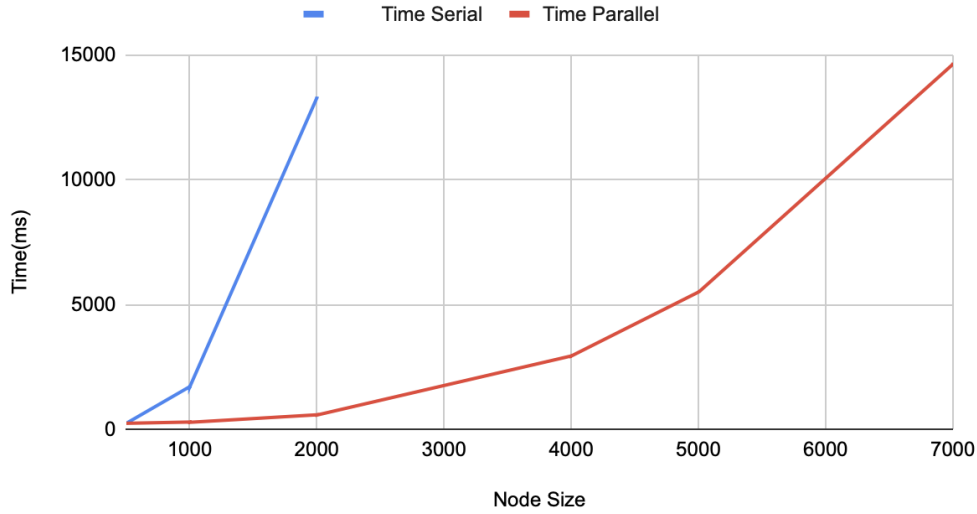


Figure 4: Comparison between the serial and parallel run times

5 Analysis of Average distance when $P=0$

In this section, we will connect to the 12 nearest nodes and we set the rewire probability to 0 in order to test our math that the average distance scales linearly with the number of nodes in the graph. We see that it does.

We see there is no effect on correctness when the code is ran in serial vs parallel which is good.

6 Analysis of average distance changing P

Here we vary the rewire probability of any given edge. There is no reason to run the code both serially and in parallel so we just run in parallel. Parameters are node size= 4000 and connect to the 12 nearest neighbors.

7 Conclusion

In this report we showed that CArrayKokkos is able to utilize parallelism effectively dramatically reducing the runtime and making larger problems feasible. Furthermore we validated the already known behavior of this random graph model. This model gives us reason to believe real world networks such as the facebook friendship network should have a surprisingly low average distance even though most of our friends are local i.e. most peoples friends are mutual friends more frequently than random chance would suggest. Finally, we note that the algorithms in this paper are easily parallelizable and converting from serial to parallel with the exception of random number generation is straightforward.

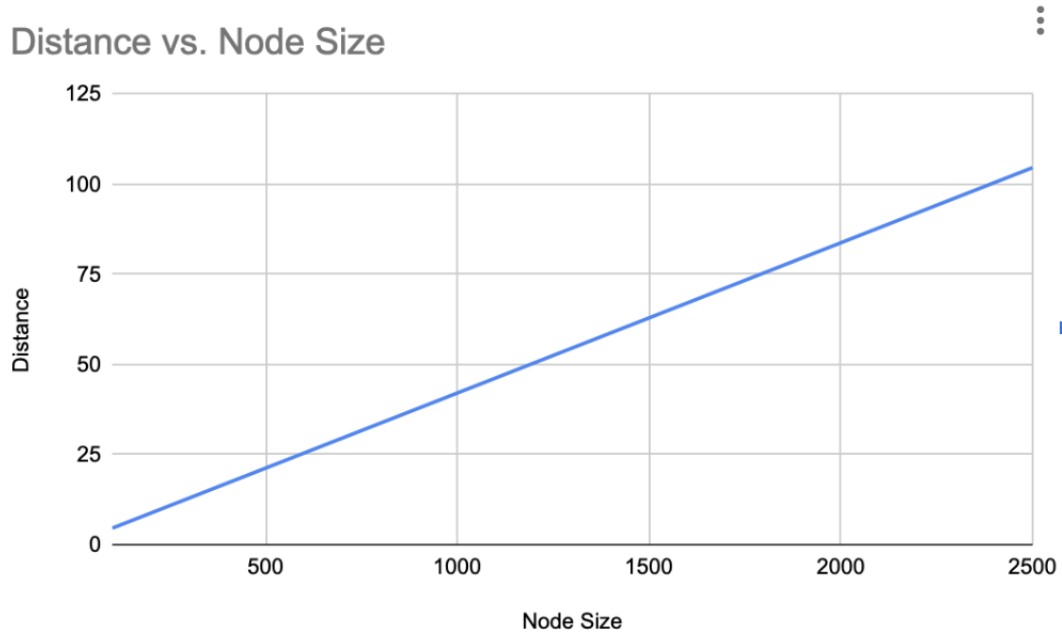


Figure 5: Average distance when ran in serial

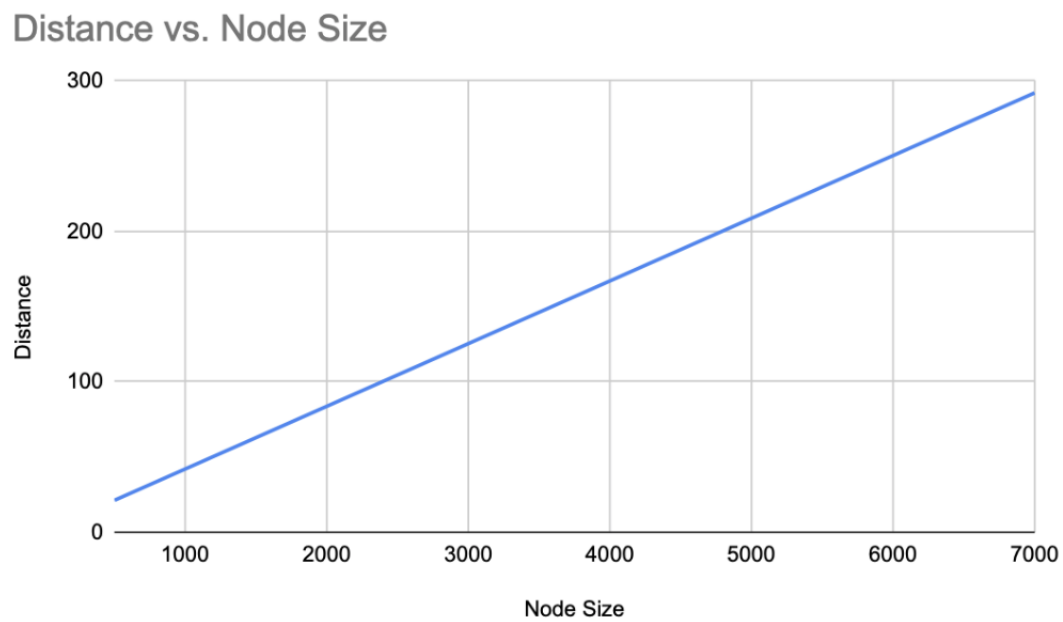


Figure 6: Average distance when ran in parallel

Probability and Distance

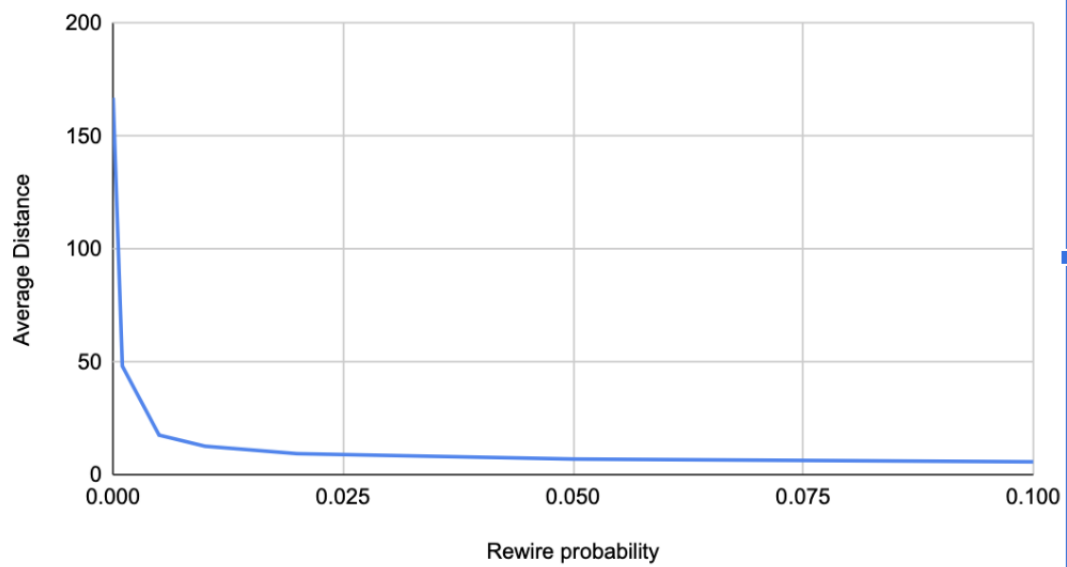


Figure 7: Affect of changing the rewire probability on average distance in the network. The effect even for small probabilities is quite strong