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Spin models on random bipartite graphs

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Abstract

This is my abstract.

Contents

1   Introduction

This project is centered around doing Ising Model simulations on random bipartite graphs. As such, it is useful to know more the Ising model and graph theory before we get started.

1.1   The Ising Model

1.1.1   What is the Ising Model?

The Ising Model is a mathematical model that was invented by Wilhelm Lenz, but developed by Ernst Ising in 1925. It is used to to explore ferromagnetism in statistical physics.

1.1.2   Analytical tools used for studying the Ising Model

1.1.3   How it is normally studied in serial and parallel simulations

The two dimenstional Ising model is usually represented as a rectangular lattice, sometimes with periodic boundary conditions and sometimes not, depending on the case in question. Each point on the lattice is assigned a spin at the start of the simulation, usually at random. Each point only 'feels' the interaction of its nearest neighbours (each point has four neighbours). These interactions allow a Hamiltonian to be calculated for each point.

FILL ABOUT METROPOLIS HASTINGS AND UPDATING POINTS

In parallel we divide the grid up into smaller subgrids. In the figure, the grid is divided into four subgrids but obviously more subdivisions are possible. Normally we have one subgrid per processor that we intend to run the program on. The need for *Message Passing Interface* (MPI) function calls arises from the fact that on the edges of our subgrids there are points needed for the update step whose values are stored on another processor. These values need to be identified and passed to the relevant processor. MPI is used for this task.

1.2   Graph Theory

Graph theory refers to the mathematical study of *graphs*. A graph is a visual representation of set of objects, known as *vertices*. Some pairs of these obects are then connected by links, known as *edges*. If the edges are said to have orientation (if edge $(a, b) \neq (b, a)$, where a, b are vertices), then we call the graph a directed graph. If the
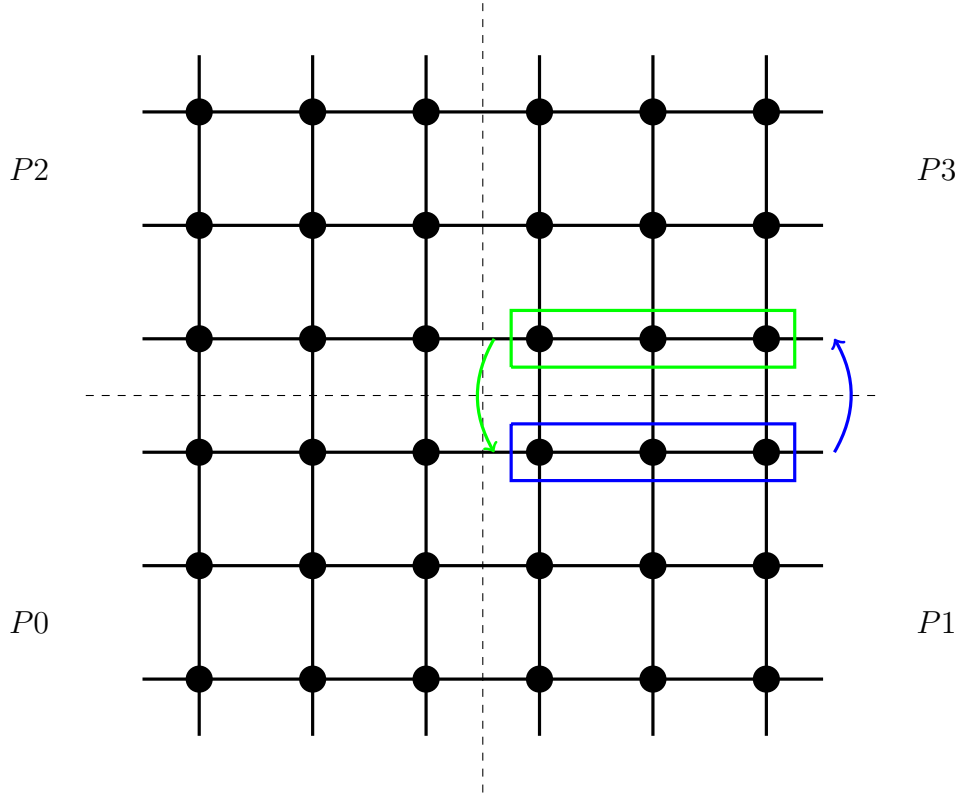
Figure 1: Regular square lattice Ising model.

edges don't have orientation (if $(a, b) = (b, a)$) then we call the graph an undirected graph. We will deal only with undirected graphs in this report.

For this project we're not going to consider disconnected graphs. That is, graphs where there are no nodes connecting a vertex, or set of connected vertices, to the rest of the graph. Only connected graphs are considered.

All graphs considered will be random, *bipartite* graphs. A bipartite graph is a graph in which we can divide its vertices into two disjoint sets, A and B, such that vertices in A are only connected to vertices in B, and vice versa. Disjoint means that the two sets have no element in common.

A random graph is a graph where the edges connect vertices at random; there is no pattern or order to how vertices are connected.

Trivalent graphs are another class of the graphs we will be dealing with a lot. For a graph to be trivalent it means that ever vertex has exactly three edges connecting it to three other distinct vertices.
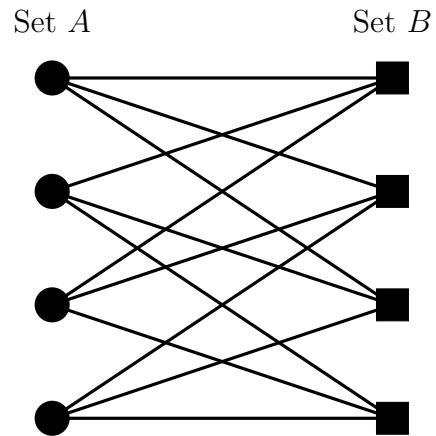
Figure 2: A random bipartite trivalent graph.

1.3 Motivation

1.3.1 Why this is an interesting MPI problem

1.3.2 Physical uses of these simulations

1.4 Aims

2 Software Architecture

In this section I will discuss the various techniques used in creating the simulations that were run in the duration of this project. The main areas these fall under are: graph generating - in both serial and parallel, as well as associated swap algorithms; MPI communication structure; and, finally, how the update step worked.

2.1 Random graph generation

Over the course of this project numerous different methods were implemented and compared for the 'best' way to create a random bipartite graph. These methods were written in both serial and parallel.

2.1.1 Generating random graphs in serial

2.1.2 Swap algorithms in serial and parallel

The swap aglorithm is used for two purposes in this project. The first in the generation of random graphs, and the second is as a proposal step during an update (instead of proposing a spin flip a swap is proposed instead. The reasoning and implementation of both of this will be addressed in a while but first we will look at how the swap alogrithm works.

The algorithm

The algorithm works by first choosing a node in set A, at random, which we will call a_0 . From this node two of its neighbours in set B are chosen (again, at random), we call these b_1 and b_2. The next step is to choose one neighbour each of b_1 and b_2, ensuring not to choose a_0 again. Label the neighbour of b_1 as a_1 and the neighbour of b_2 as a_2. At this point we have the nodes selected that we wish to swap. We want to have b_1 not connected to a_1 anymore, but instead have it connected to a_2, and similarly no longer have b_2 connected to a_2, but connected to a_1.

Before we can do this however we must preform some checks. We must ensure that a_1 is not connected to b_2 by any other edge that we have not considered so far, and similarly we must must ensure that a_2 is in no way connected to b_1. The reason for this check is that we may already have the nodes we wish to swap connected to each other by another edge, so if the swap is performed then it would mean having two edges connecting the same two points to each other, which is not desirable.
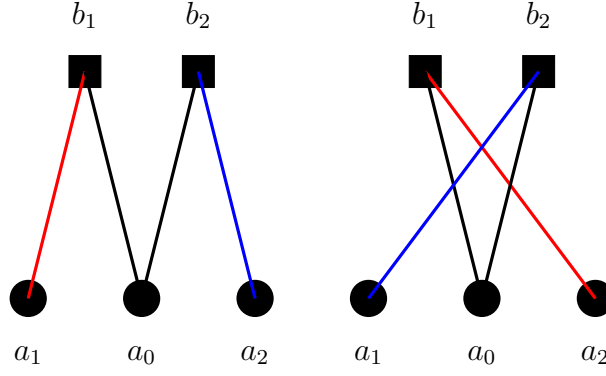
Figure 3: Swap alg.

Serial Implementation

Implementing the algorithm in serial is a rather straightforward. Pseudocode of thel

Parallel Implementation

The parallel implementation of this algorithm is, naturally, a little more tricky. It is also a problem that does not parallise well at all. In fact, the serial algorithm is much faster and more efficient.

2.2 MPI communications

2.2.1 Data division and load balancing

The problem of how to split the data most efficiently between computing cores will be handled in this section. When we want to solve the system in parallel we have to give each processor a certain number of points on the graph. The method used to divide the data was to simply divide the nodes equally among the processors.

Algorithm 1 Serial swap algorithm
___
Pick $a_0 = \mathtt{rand} \in A$

while $b_1 \neq b_2$ **do**

    $b_1 = \mathtt{rand} \in N(a_0)$

    $b_2 = \mathtt{rand} \in N(a_0)$

end while

while $flag = 1$ **do**

    while $a_1 \neq a_2$, $a_1 \neq a_0$, $a_2 \neq a_0$ **do**

        $a_1 = \mathtt{rand} \in N(b_1)$

        $a_2 = \mathtt{rand} \in N(b_2)$

    end while

    if $a_1 \in N(b_2)$ **then**

        $flag = 1$

    end if

    if $a_2 \in N(b_1)$ **then**

        $flag = 1$
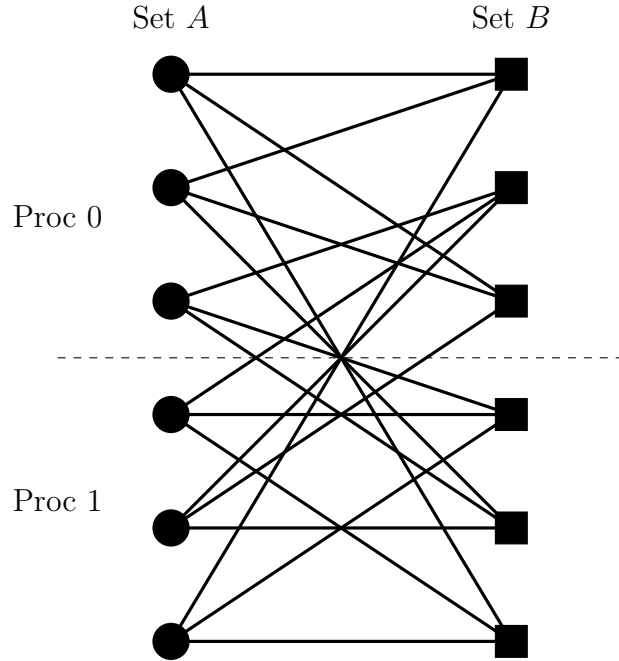
    end if

end while
___

Figure 4: Data division for a random trivalent, bipartite graph.

Algorithm 2 Parallel swap algorithm
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  if $rank = 0$ **then**

      Pick $a_0 = \texttt{rand} \in A$

      Determine rank of proc a_0 is hosted on, $rank(a_0)$

  end if

  `MPI_Bcast(a_0,...,0,...)`

  `MPI_Bcast(rank(a_0),...,0,...)`

  if $rank = rank(a_0)$ **then**

      while $b_1 \neq b_2$ **do**

         $b_1 = \texttt{rand} \in N(a_0)$

         $b_2 = \texttt{rand} \in N(a_0)$

      end while

  end if

  `MPI_Bcast(b_1,...,rank(a_0),...)`

  `MPI_Bcast(b_2,...,rank(a_0),...)`

  Determine rank of proc b_1 is hosted on, $rank(b_1)$

  Determine rank of proc b_2 is hosted on, $rank(b_2)$

  while $a_1 \neq a_2$, $a_1 \neq a_0$, $a_2 \neq a_0$ **do**

      if $rank = rank(b_1)$ **then**

         $a_1 = \texttt{rand} \in N(b_1)$

      end if

      if $rank = rank(b_2)$ **then**

         $a_2 = \texttt{rand} \in N(b_2)$

      end if

  end while

  `MPI_Bcast(a_1,...,rank(b_1),...)`

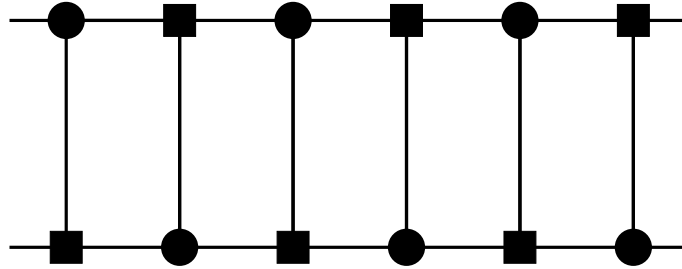 `MPI_Bcast(a_2,...,rank(b_2),...)`

Figure 5: Double ring bipartite graph.

3   Results

4 Conclusion

hello

5 Future Work