TWO NETWORKS DERIVED FROM THE HYPERCUBE

by

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TWO NETWORKS DERIVED FROM THE

HYPERCUBE

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Abstract

The hypercube is one of the most popular interconnection networks. Not only does it have good topological structure but also nice symmetric properties. However, it has a major drawback that the degree is not bounded as the dimension increases. Because of this, some networks with bounded degree have been derived from the hypercube. Two of the most popular are butterfly graphs and cube-connected-cycles. They both inherit some properties from the hypercube. This thesis investigates these two networks.
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Chapter 1

Introduction

Networks can be divided into two classes: static networks and dynamic networks. The physical structure of dynamic networks is not fixed. It can be changed by modifying the configuration of the switches in the connection cables. On the other hand, the physical structure of static networks is fixed. It can be modelled by the tools from Graph Theory. In this thesis, only static networks are considered.

Before constructing a network, one has to consider many factors such as the hardware cost, the performance, the reliability and the expandibility. The first three are obviously important factors. For the expandibility, the network must be designed to minimize the total amount of modification when more processors are being added to it. The simplest method of building a network is by putting a link between a new processor and one of the processors randomly chosen in the network. The cost for this kind of network is very low because not many links are required. The expandibility is clearly very high because almost no modification is required when a new processor is added. However, this kind of connection is unpredicable. The network may turn out to be a simple path. That means if any one of the links or one of the processors breaks down, the network will not be connected. Furthermore, the time for communication is very long for the processors at the end of the path. Thus, this kind of network in general is neither reliable nor efficient. Another extreme method is to connect every pair of processors. This kind of network has the maximum performance and reliability. Each processor in the network can directly communicate with every other. Also, the network is
always connected no matter how many processors are out-of-function. However, the cost will be very high. The expandibility is also not very good because the number of links going out from each processors is not the same for the networks of different size. Thus, every processor must be modified if the network need to be expanded. These two examples show that building a network is a trade-off problem.

For the hardware cost, one should consider the number of links being used. In order to increase the performance, the network should be designed so that any pair of processors can communicate easily. One measure of this is that the diameter of the network should be small.

The design of a network also affects the software cost. If processors can be addressed using binary numbers, the system software will be simpler because most of the other components in the whole system are binary-based. Furthermore, if the network is exactly "the same" with respect to each processor, the processors can share a single routing table. The network is said to have good symmetry properties. Symmetry properties also affect the reliability of the network. Good symmetric network can re-order the processors or communication lines so that some particular processors can still communicate.

This thesis will present two methods for constructing a network that has good symmetry properties. One of them is the Cayley graph construction and the other one is the set graph construction. This thesis will mainly discuss the Cayley graph construction.

The \(n\)-dimensional (binary) hypercube is one of the most popular networks that is formed by the Cayley graph construction. It does not have too many links but it has very good performance. The routing in the hypercube is extremely easy. The reliability is also very high. Moreover, it possesses all the symmetry properties that one usually studies. Unfortunately, it has one major drawback in that the number of links going out from the processors increases as the dimension increases. It reduces the expandibility dramatically.

There are some extensions coming from the hypercube. The two popular ones are the butterfly graph and the cube-connected-cycle. Both of them use an \(n\)-cycle to replace each processor in the the hypercube so that the degree of each processor can be fixed. They inherit many of the topological properties from the hypercube, but they also destroy many of the symmetry
properties. This means the performance of the butterfly graph and the cube-connected-cycle will be almost the same as the hypercube, but they have lower reliability. This thesis will investigate these two networks, and will discuss the topological structure as well as the symmetry properties of the butterfly graph and the cube-connected-cycles. It also does some comparisons between these two kinds of networks with the hypercube.
Chapter 2

Groups of Permutations

2.1 Permutation Groups

Permutation groups play a very important role in group theory. In fact, Cayley’s Theorem [5] says that every finite group is isomorphic to a group of permutations. In this section, certain notions regarding permutation groups will be presented.

Given the set \(\{1, \ldots, n\}\), one can think of a permutation as a rearrangement of the numbers. The following is a formal definition of a permutation [5].

**Definition 2.1** A function \(f : A \rightarrow B\) is one-to-one if every element of \(B\) has at most one element of \(A\) mapped to it.

**Definition 2.2** A function \(f : A \rightarrow B\) is onto if every element of \(B\) has at least one element of \(A\) mapped to it.

**Definition 2.3** A permutation of a set \(A\) is a one-to-one and onto function from \(A\) to \(A\).

In this chapter, all permutations are on the set \(\{1, \ldots, n\}\). The collection of all permutations of \(\{1, \ldots, n\}\) is usually denoted by \(S_n\).
A permutation, \( \sigma \in S_n \) can be written in several ways. One of them is to list all \((x, \sigma(x))\) pairs in an array as:

\[
\sigma = \begin{pmatrix}
1 & 2 & \ldots & i & \ldots & n \\
\sigma(1) & \sigma(2) & \ldots & \sigma(i) & \ldots & \sigma(n)
\end{pmatrix}.
\]

In fact, all permutations in \( S_n \) have the same first row. This row is actually redundant. Hence, \( \sigma \) can be written as:

\[
\sigma = \sigma(1)\sigma(2)\ldots\sigma(i)\sigma(n)
\]

**Definition 2.4** Let \( \rho, \sigma \in S_n \). A binary operation \( \cdot \) is defined as the composition of functions, that is, \((\rho \cdot \sigma)(x) = \rho(\sigma(x))\) for all \( x \in \{1, \ldots, n\} \).

**Proposition 2.5** \( S_n \) is closed under \( \cdot \).

*Proof:* Let \( \rho, \sigma \in S_n \). For any \( x, y \in \{1, \ldots, n\} \), \((\rho \cdot \sigma)(x) = (\rho \cdot \sigma)(y) \Rightarrow \rho(\sigma(x)) = \rho(\sigma(y)) \). Since \( \rho \) is one-to-one, \( \sigma(x) = \sigma(y) \). Again, \( \sigma \) is one-to-one implying that \( x = y \). Hence, \( \rho \cdot \sigma \) is one-to-one.

Now for any \( y \in \{1, \ldots, n\} \), there is \( x \in \{1, \ldots, n\} \) such that \( y = \rho(x) \). Again, there is \( z \in \{1, \ldots, n\} \) such that \( x = \sigma(z) \). Hence, \( y = \rho(\sigma(z)) = (\rho \cdot \sigma)(z) \) and \( \rho \cdot \sigma \) is onto. Thus \( \rho \cdot \sigma \) is a permutation in \( S_n \). \( \square \)

**Proposition 2.6** \((S_n, \cdot)\) is a group.

*Proof:* Let \( e \) be the permutation such that \( e(x) = x \) for all \( x \in \{1, \ldots, n\} \). Then for any \( \rho \in S_n \), \((\rho \cdot e)(x) = \rho(e(x)) = \rho(x)\) for all \( x \in \{1, \ldots, n\} \). Hence, \( e \) is an identity.

For any \( \rho, \sigma, \tau \in S_n \), \[((\rho \cdot \sigma) \cdot \tau)(x) = (\rho \cdot (\sigma \cdot \tau))(x) = (\rho(\sigma(\tau(x)))) = \rho((\sigma \cdot \tau)(x)) = (\rho \cdot (\sigma \cdot \tau))(x)\) for all \( x \in \{1, \ldots, n\} \). Hence, \( \cdot \) is associative.

Let \( \rho \) be a permutation. As \( \rho \) is one-to-one and onto, the inverse function \( \rho^{-1} \) of \( \rho \) exists, and \( \rho^{-1} \) is also one-to-one and onto. We have \((\rho \cdot \rho^{-1})(x) = x = e(x)\) for all \( x \in \{1, \ldots, n\} \). Hence, \((S_n, \cdot)\) is a group. \( \square \)

\((S_n, \cdot)\) is usually called the symmetric group of degree \( n \).
2.2 Orbits and Cycles

Given a permutation \( \sigma \), one can partition \( \{1, \ldots, n\} \) using an appropriate relation \( \sim \) defined as follows: For any \( a, b \in \{1, \ldots, n\} \), \( a \sim b \) if and only if \( b = \sigma^n(a) \) for some integer \( n \).

**Proposition 2.7** The relation \( \sim \) is an equivalence relation.

**Proof:**

**Reflexive** \( a \sim a \) because \( a = e(a) = \sigma^0(a) \).

**Symmetric** \( a \sim b \Rightarrow b = \sigma^n(a) \) for some integer \( n \). So \( a = \sigma^{-n}(b) \) and \( b \sim a \).

**Transitive** \( a \sim b, b \sim c \Rightarrow b = \sigma^n(a), c = \sigma^m(b) \) for some integers \( m \) and \( n \).

So \( c = \sigma^m(\sigma^n(a)) = \sigma^{m+n}(a) \) and \( a \sim c \). \( \Box \)

**Definition 2.8** Let \( \sigma \in S_n \), the equivalence classes determined by \( \sim \) are called the *orbits* of \( \sigma \).

Hence a permutation partitions the set \( \{1, \ldots, n\} \) into orbits. This idea provides a method to decompose a permutation into a set of simple permutations.

Another way to describe a permutation is to use a digraph. Let \( \sigma \in S_n \) and \( V = \{1, \ldots n\} \) be the vertex-set. There is an arc from \( i \) to \( j \) if and only if \( \sigma(i) = j \). We denote this digraph by \( D_\sigma \). Clearly, a directed cycle in \( D_\sigma \) corresponds to an orbit in \( \sigma \).

**Theorem 2.9** The associated digraph \( D_\sigma \) of \( \sigma \) consists of a set of vertex-disjoint directed cycles.

**Proof:** As \( \sigma \) is a function, the outdegree of each vertex in \( D_\sigma \) is 1. Since \( \sigma \) is one-to-one and onto, the indegree of each vertex in \( D_\sigma \) is also 1. Hence, \( D_\sigma \) consists of a set of vertex-disjoint directed cycles. \( \Box \)
Theorem 2.9 says that we can decompose $D_\sigma$ into a set of vertex-disjoint directed cycles, $C_1, C_2, \ldots, C_k$. Let $B_1, B_2, \ldots, B_k$ be the directed spanning subgraphs induced by $C_1, C_2, \ldots, C_k$, respectively, and add a directed loop to each isolated vertex. Each $B_i$ will give us a permutation. Those permutations have at most one orbit containing more than one element. Since the permutations come from the cycles of $D_\sigma$, these permutations are also named cycles.

**Definition 2.10** A permutation $\sigma \in S_n$ is a *cycle* if $\sigma$ has at most one orbit containing more than one element. The *length* of a cycle is the number of elements in the largest orbit.

**Definition 2.11** Two cycles are said to be *disjoint* if their orbits that contain more than one element do not have any element in common.

**Corollary 2.12** Every permutation $\sigma \in S_n$ can be written as a product of disjoint cycles.

### 2.3 Cyclic Notation

By the definition, a cycle has at most one orbit containing more than one element. So given the list of the elements in the largest orbit of a cycle is sufficient to determine the whole structure of the permutation. For example, if the largest orbit of a cycle $c \in S_8$ is $(1 8 3 5)$, then

$$c = \begin{pmatrix} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \\ 8 & 2 & 5 & 4 & 1 & 6 & 7 & 3 \end{pmatrix}.$$  

The notation, $c = (1 8 3 5)$ is called *cyclic notation*.

Corollary 2.12 says that every permutation can be written as a product of disjoint cycles. For example $p = 6734152$ can be written as $(1 6 5)(2 7)$.
2.4 Transpositions and Inversions

Other than the identity, every cycle has length at least 2. This means that a cycle of length 2 has the simplest structure. However, it has a special property. In cyclic notation, a cycle of length 2 can be written as \((i \ j)\). In explicit notation, it will look like \(1 \ldots i-1 \ i \ i+1 \ldots j-1 \ i \ j \ i+1 \ldots n\). Given a cycle \(p\) of length 2 and a permutation \(\sigma = a_1 a_2 \ldots a_n\) in explicit notation. If \(p\) is multiplied on the right of \(\sigma\), it will exchange the \(i\)th and \(j\)th elements in \(\sigma\). Similarly if \(p\) is multiplied on the left of \(\sigma\), it will exchange \(i\) and \(j\) in \(\sigma\). Hence, a cycle of length 2 is called a transposition. Using this property, one can obtain any permutation by exchanging suitable pairs of elements.

**Theorem 2.13** Every permutation in \(S_n, n \geq 2\), can be written as a product of transpositions.

**Proof:** It is sufficient to prove that every cycle can be written as a product of transpositions. For any cycle \(c = (a_1 \ a_2 \cdots a_k)\) in cyclic notation, \(k \geq 2\), it can be written as

\[
c = (a_1 \ a_k)(a_1 \ a_{k-1})\cdots(a_1 \ a_2).
\]

If the length of \(c\) is 1, then \(c\) is the identity and can be written as \((1\ 2)(1\ 2)\).

\( \qed \)

Another property of a transposition is that every transposition is the inverse of itself. Suppose \(\sigma\) can be written as a product \(p_1 p_2 \cdots p_k\) of \(k\) transpositions. Then \(e = \sigma p_k p_{k-1} \cdots p_1\), where \(e\) is the identity. This is the basic idea of a sorting algorithm.

**Definition 2.14** Let \(\sigma\) be a permutation. If \(\sigma(j) < \sigma(i)\) for \(i < j\), then the pair \((\sigma(j), \sigma(i))\) is called an inversion of \(\sigma\).

Transpositions and inversions in fact are the crucial parts of the sorting algorithm based on comparison. The rest of this section will discuss the relationship between transpositions and inversions.

**Definition 2.15** A permutation \(\sigma\) is said to be an odd or even permutation if the number of inversions in \(\sigma\) is odd or even, respectively.
Lemma 2.16 Let $w = a_1a_2 \cdots a_n$ be a permutation and $p = 1 \ldots i - 1 j i + 1 \ldots j - 1 i j + 1 \ldots n$ be a transposition. The number of inversions in $w$ and in $wp$ has different parity.

Proof: We have $wp = a_1a_2 \cdots a_{i-1}a_ia_{i+1} \cdots a_{j-1}a_ia_{j+1} \cdots a_n$. It is sufficient to consider the subsequence $a_ia_{i+1} \cdots a_{j-1}a_i$.

If $a_i < a_j$, then $(a_i, a_j)$ is an inversion in $wp$ but not in $w$. For any $a_i < a_k < a_j$, $i + 1 \leq k \leq j - 1$, both $(a_k, a_j)$ and $(a_i, a_k)$ are inversions in $wp$ but not in $w$. If $a_i < a_j < a_k$ or $a_k < a_i < a_j$, $i + 1 \leq k \leq j - 1$, then the number of inversions involving $a_k$ in $wp$ is the same as in $w$. Thus, the number of inversions in $wp$ is increased by an odd number.

If $a_i > a_j$, then $(a_j, a_i)$ is an inversion in $w$ but not in $wp$. For any $a_i > a_k > a_j$, $i + 1 \leq k \leq j - 1$, both $(a_k, a_i)$ and $(a_j, a_k)$ are inversions in $w$ but not in $wp$. If $a_i > a_j > a_k$ or $a_k > a_i > a_j$, $i + 1 \leq k \leq j - 1$, then the number of inversions involving $a_k$ in $w$ is the same as in $wp$. Thus, the number of inversions in $wp$ is decreased by an odd number. The result follows.

Theorem 2.17 Let $w = p_1p_2 \cdots p_k$ be a product of $k$ transpositions. Then $w$ is even (or odd) if and only if $k$ is even (or odd).

Proof: Let $e$ be the identity. The number of inversions in $e$ is 0. When $k = 1$, $k$ is odd and $w = p_1 = ep_1$. By Lemma 2.16, $w$ has an odd number of inversions, that is, $w$ is odd. Suppose it is true for $k = r - 1$. Consider $k = r$. By Lemma 2.16, $w = p_1 \cdots p_{r-1}p_r$ and $p_1 \cdots p_{r-1}$ have different parity. The result follows.

Corollary 2.18 If $\sigma$ is even (or odd), then $\sigma$ can only be written as a product of an even (or odd) number of transpositions.

Proof: By Theorem 2.13, $\sigma$ is a product of transpositions. Let $\sigma = p_1p_2 \cdots p_k$, where the $p_i$'s are transpositions. By Theorem 2.17, the result follows.
2.5 Conjugacy

We again consider the associated directed graph $D_\sigma$ of the permutation $\sigma$ again. $D_\sigma$ consists of a set of vertex-disjoint cycles. The length of the cycles can be any number from 1 to $n$, so let $[\lambda_1(\sigma), \lambda_2(\sigma), \ldots \lambda_n(\sigma)]$ be an $n$-tuple, where $\lambda_i(\sigma)$ is the number of cycles of length $i$. We define a relation $\sim_c$ as follows: For any $\sigma_1, \sigma_2 \in S_n$, $\sigma_1 \sim_c \sigma_2$ if and only if

$$[\lambda_1(\sigma_1), \lambda_2(\sigma_1), \ldots, \lambda_n(\sigma_1)] = [\lambda_1(\sigma_2), \lambda_2(\sigma_2), \ldots, \lambda_n(\sigma_2)].$$

Definition 2.19 The relation $\sim_c$ is called conjugacy.

Theorem 2.20 Conjugacy is an equivalence relation.

Proof: The proof is trivial.

2.6 Stabilizer

There are special subgroups in a permutation group which we now define.

Definition 2.21 Let $(B, \cdot)$ be a permutation group. Let $u \in \{1, \ldots, n\}$ and $B_u = \{\alpha : \alpha \in B \text{ and } \alpha(u) = u\}$. $B_u$ is called the stabilizer of $u$.

Proposition 2.22 We have that $(B_u, \cdot)$ is a subgroup of $(B, \cdot)$.

Proof: Let $\alpha, \beta \in B_u$. Then $(\alpha \cdot \beta)(u) = \alpha(\beta(u)) = \alpha(u) = u$. So $\alpha \cdot \beta \in B_u$. Since $e(u) = u$, $e \in B_u$ If $\alpha \in B_u$, $\alpha(u) = u$. So $\alpha^{-1}(u) = u$, that is, $\alpha^{-1} \in B_u$. Hence, $(B_u, \cdot)$ is a subgroup of $(B, \cdot)$. \qed
2.7 Transitive groups and regular groups

There are certain permutation groups that play important roles in group theory. Some of them will be used in the coming chapters.

Definition 2.23 Let $\Gamma$ be a permutation group on $\{1, \ldots, n\}$. $\Gamma$ is transitive if for each pair $i, j \in \{1, \ldots, n\}$, there exists a $\sigma \in \Gamma$ such that $\sigma(i) = j$.

Definition 2.24 A permutation group $\Gamma$ on $\{1, \ldots, n\}$ is said to be regular if $\Gamma$ is transitive and for each $i \in \{1, \ldots, n\}$, the stabilizer $\Gamma_i$ of $i$ is $\{e\}$.

Theorem 2.25 A permutation group $\Gamma$ on $\{1, \ldots, n\}$ is regular if and only if for any pair $i, j \in \{1, \ldots, n\}$, there is a unique permutation $\sigma \in \Gamma$ such that $\sigma(i) = j$.

Proof: Since $\Gamma$ is transitive, it is sufficient to show that $\sigma$ is unique. Suppose there are two permutations $\sigma_1$ and $\sigma_2$ such that $\sigma_1(i) = \sigma_2(i) = j$. Then $e(i) = i = \sigma_1^{-1}\sigma_1(i) = \sigma_1^{-1}\sigma_2(i)$. Therefore, $\sigma_1^{-1}\sigma_1 = \sigma_1^{-1}\sigma_2$ implies that $\sigma_1 = \sigma_2$. \qed
Chapter 3

Cayley Graphs and Transposition Graphs

3.1 Cayley Graphs

Cayley graphs are an important class of graphs constructed from groups. They reflect not only the group structure but also possess some nice graph properties.

Definition 3.1 Let \((\Gamma, \cdot)\) be a finite group with identity \(e\). Let \(S\) be a subset of \(\Gamma\) such that

1. if \(g \in S\), then \(g^{-1} \in S\), and
2. \(e \notin S\).

The Cayley graph \(G(\Gamma, S)\) is defined as follows.

1. The vertex set of \(G(\Gamma, S) = \Gamma\).
2. The edge set of \(G(\Gamma, S) = \{xy : x, y \in \Gamma\) and there exists \(g \in S\) such that \(y = x \cdot g\}\).

The set \(S\) is called the symbol of \(G(\Gamma, S)\).

Proposition 3.2 \(S\) generates \(\Gamma\) if and only if \(G(\Gamma, S)\) is connected.
Proof: Let \( S = \{a_1, \ldots, a_r\} \). Suppose \( S \) generates \( \Gamma \). Let \( x, y \in G(\Gamma, S) \) and \( g = x^{-1}y \). Since \( \Gamma = \langle S \rangle \), \( g = x^{-1}y = a_i a_i \cdots a_i \). Hence, \( y = xx^{-1}y = xa_i a_i \cdots a_i \). This implies \( x, y \) are connected by a path. Conversely, suppose \( G(\Gamma, S) \) is connected. Let \( g \in \Gamma \). There is a path from \( e \) to \( g \) in \( G(\Gamma, S) \). So \( g = a_i a_i \cdots a_i \). This implies \( S \) generates \( \Gamma \).

Since there is no reason to consider a disconnected interconnection network, all symbol sets in this thesis will be assumed to be generator sets.

3.2 Transposition Graphs

Another kind of graph that is determined by a permutation group is a transposition graph. Studying transposition graphs is not useful because every simple graph is a transposition graph. However, the transposition graph corresponding to the Cayley graph generated by a permutation group has some special characteristics.

Definition 3.3 Let \( (\Gamma, \cdot) \) be a permutation group on \( A \). Let \( S \) be a set of transpositions in \( \Gamma \). The transposition graph \( TG(A, S) \) is defined as follows:

1. The vertex set of \( TG(A, S) = A \), and
2. The edge set of \( TG(A, S) = \{xy : (x, y) \in S\} \).

Definition 3.4 A transposition graph which is a tree is called a transposition tree.

Theorem 3.5 (Pólya) A set \( \Omega \subseteq S_n \) of \((n - 1)\) transpositions generates the symmetric group \( S_n \) if and only if the transposition graph \( TG(S_n, \Omega) \) is a transposition tree.

Proof: Suppose \( TG(S_n, \Omega) \) is a tree. Then any two vertices are connected by a unique path. Let \( a, b \in \{1, \ldots, n\} \) and

\[
a, (a \ x_1), x_1, (x_1 \ x_2), \ldots, (x_{k-1} \ x_k), x_k, (x_k \ b), b
\]
be the path joining \( a \) and \( b \). Then

\[(a \ b) = (a \ x_1) (x_1 \ x_2) \cdots (x_k \ x_k) (x_k \ b) (x_{k-1} \ x_k) \cdots (x_1 \ x_2) (a \ x_1)\]

which is a product of transpositions in \( \Omega \).

Conversely, suppose \( \Omega \) generates \( S_n \). Let \( (x \ y) = p_1 p_2 \cdots p_k \), where \( p_1, p_2, \ldots, p_k \in \Omega \). Then \( p_{i_1} = (x \ x_1) \) for some \( x_1 \) and \( 1 \leq i_1 \leq k \). Similarly, \( p_{i_2} = (x_1 \ x_2) \) for some \( x_2 \) and \( 1 \leq i_2 \leq k \), \( p_{i_3} = (x_2 \ x_3) \) for some \( x_3 \) and \( 1 \leq i_3 \leq k \), and so forth. Finally, \( p_{i_r} = (x_{r-1} \ y) \) for some \( 1 \leq i_r \leq k \). Clearly \( x \) and \( y \) are joined by a walk \( x, x_1, x_2, \ldots, x_{r-1}, y \) in \( TG(S_n, \Omega) \). Hence, \( TG(S_n, \Omega) \) is connected with \( n - 1 \) edges, that is, it is a tree.

\[\square\]

**Corollary 3.6** A set \( \Omega \subseteq S_n \) of transpositions generates the symmetric group \( S_n \) if and only if the transposition graph \( TG(S_n, \Omega) \) is connected.

**Proof:** Every connected graph has a spanning tree. By Theorem 3.5, the result follows. \[\square\]
Chapter 4
Symmetry in Graphs

Symmetry is an important issue in interconnection networks. It affects not only the performance but also the cost of the network. For instance, if a network has symmetry on the nodes, the same routing algorithm can be used on each node. This simplifies both the hardware of the control center and the system software of the operating system. This chapter will discuss certain symmetry that a network can have.

4.1 Automorphisms on Graphs

Given a square, one can rotate it and flip it. The square is still a square. However, if one tries to “twist it”, the square will no longer be a square. On the other hand, no transformation can make a complete graph structurally different. This kind of transformation that preserves the structure of the graph is called an automorphism.

Definition 4.1 A vertex automorphism $\alpha$ of $G$ is a permutation of the vertex-set that preserves the adjacency. That is, if the edge $xy \in E$, then the edge $\alpha(x)\alpha(y) \in E$. 

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4.2 Transitivity

In a network, it will be useful if the network looks the same when viewed through any node. In other words, each node lead to the same network by relabelling the other nodes. This property is called vertex-transitivity.

**Definition 4.2** $G$ is said to be vertex-transitive if given any pair of vertices $x$ and $y$, there exists $\alpha \in Aut(G)$ such that $y = \alpha(x)$.

**Definition 4.3** $G$ is said to be edge-transitive if given any pair of edges $xy$ and $uv$, there exists $\alpha \in Aut(G)$ such that $x = \alpha(u)$ and $y = \alpha(v)$, or $x = \alpha(v)$ and $y = \alpha(u)$.

From this definition, it is easy to see that a vertex-transitive graph has to be regular because no automorphism can map a vertex to one of different degree.

Vertex-transitivity can be generalized. Let $D$ be the diameter of the graph $G$. For $0 \leq k \leq D$, $G$ is said to be $k$-distance-transitive if given four vertices, $x, y, u$ and $v$ such that $d(x, y) = d(u, v) = k$, then there exists an $\alpha \in Aut(G)$ such that $u = \alpha(x)$ and $v = \alpha(y)$. If $G$ is $k$-distance-transitive for all $0 \leq k \leq D$, then it is called distance-transitive.

Clearly, vertex-transitivity is 0-distance-transitivity.

**Proposition 4.4** If a graph $G$ is 1-distance-transitive, then $G$ is edge-transitive.

**Proof:** Suppose $G$ is 1-distance-transitive. Let $e_1 = xy$ and $e_2 = uv$ be two edges in $G$. Then there exists $\alpha \in Aut(G)$ such that $u = \alpha(x)$ and $v = \alpha(y)$. The result follows. \qed

The rest of this section will discuss transitivities of the graphs. For $u \in V(G)$, define $N_i \subseteq V(G)$ as $N_i = \{v : v \in V(G) \text{ and } d(u, v) = i\}$ and $d_i = |N_i(u)|$. Then the following is the characterization of distance-transitive graph [8, 2].
Lemma 4.5 Let $D$ be the diameter of the graph $G$. $G$ is distance-transitive if and only if it is vertex-transitive and the vertex stabilizer $A_u$ is transitive on the set $N_i(u)$ for all $i \in \{0, 1, \ldots, D\}$ and for each $u \in V(G)$.

Proof: $G$ is distance-transitive implying that $G$ is 0-distance-transitive. Thus $G$ is vertex-transitive. Let $u$ be any vertex and $p, q \in N_i(u), 0 \leq i \leq D$. Since $G$ is distance-transitive, there exists $\alpha \in \text{Aut}(G)$ such that $u = \alpha(u), p = \alpha(q)$. Since $u = \alpha(u), \alpha \in A_u(G)$. So $A_u(G)$ is transitive on $N_i(u)$.

Conversely, $G$ is vertex-transitive and $A_u(G)$ is transitive on $N_i(u)$, for all $u \in V(G)$ and $i \in \{0, \ldots, D\}$. Let $x, y, u, p \in V(G)$ so that $d(x, y) = d(u, p) = d$. Let $w \in V(G)$. There exists $\alpha \in \text{Aut}(G)$ such that $w = \alpha(x)$. Let $y' = \alpha(y)$. Also, there exists $\beta \in \text{Aut}(G)$ such that $w = \beta(u)$. Let $p' = \beta(p)$. Since $\alpha$ and $\beta$ are automorphisms, $d(w, y') = d(x, y) = d = d(u, p) = d(w, p')$. So $y', p' \in N_d(w)$. Since $A_w(G)$ is transitive on $N_d(w)$, there exists an automorphism $\tau \in A_w(G)$ such that $w = \tau(w)$ and $p' = \tau(y')$. Then $u = \beta^{-1} \tau \alpha(x)$ and $p = \beta^{-1} \tau \alpha(y)$. Thus, $G$ is distance-transitive. \qed

4.3 Intersection Number

Lemma 4.5 provides a method to check whether a graph is distance-transitive. The procedure in fact is quite tedious. There is a necessary condition for a graph being distance-transitive [8, 2]. First we define $n_{hi}(u, v) = |\{w : w \in V, d(u, w) = h \text{ and } d(v, w) = i\}|$. If a graph $G$ is distance-transitive, $n_{hi}(u, v)$ is independent of $u$ and $v$ but depends only on $j$ which is $d(u, v)$. This means $n_{hi}(u, v)$ can be denoted as $n_{hij}$.

Definition 4.6 Let $D$ be the diameter of the distance-transitive graph $G$. The $(D + 1)^3$ integers $n_{hij}$ for $0 \leq h, i, j \leq D$ are called intersection numbers.

Proposition 4.7 We have $n_{1ij} = 0$ for $i \notin \{j - 1, j, j + 1\}$.

Proof: As the graph is distance-transitive, it is sufficient to consider one pair $u$ and $v$ of vertices with distance $j$ between them. Let $w$ be a vertex that is adjacent to $u$. Then $d(u, w) = 1$. Since $d(u, v) = j$, we have $j - 1 \leq d(v, w) \leq j + 1$. In other words, $n_{1ij} = 0$ for $i \notin \{j - 1, j, j + 1\}$. \qed
For this reason, if $D$ is the diameter of a distance-transitive graph, we can let

\[
a_j = n_{1jj} = |N_1(u) \cap N_j(v)|
b_j = n_{1,j+1,j} = |N_1(u) \cap N_{j+1}(v)|
c_j = n_{1,j-1,j} = |N_1(u) \cap N_{j-1}(v)|,
\]

where $u$ and $v$ are any pair of vertices with distance $j$ between them, $0 \leq j \leq D$. Furthermore, $b_D$ and $c_0$ are undefined. These $3D + 1$ integers can be arranged as an array.

**Definition 4.8** The array

\[
IA(G) = \begin{bmatrix}
a_0 & a_1 & a_2 & \ldots & a_{D-1} & a_D \\
b_0 & b_1 & b_2 & \ldots & b_{D-1} & * \\
* & c_1 & c_2 & \ldots & c_{D-1} & c_D
\end{bmatrix}
\]

is called the *intersection array* of the distance-transitive graph $G$.

The intersection array has the following properties [2].

**Lemma 4.9** If $G$ is distance-transitive, then the entries of $IA(G)$ satisfy:

1. $a_0 = 0$, $b_0 = d_1$, $c_1 = 1$,
2. $c_i + a_i + b_i = d_i$ for all $1 \leq i \leq D - 1$,
3. $1 \leq c_2 \leq c_3 \leq \ldots \leq c_D$,
4. $d_1 \geq b_1 \geq b_2 \geq \ldots \geq b_{D-1}$,
5. $d_{i-1}b_{i-1} = d_ic_i$ for $1 \leq i \leq D$,

where $d_i = |N_i(u)|$ and $D = \text{diameter}$.

**Proof:**

1. We have $a_0 = n_{100} = |N_1(u) \cap N_0(v)| = 0$ as $d(u, v) = 0$, that is, $u = v$. Also, $b_0 = n_{110} = |N_1(u) \cap N_1(v)| = |N_1(u)| = d_1$ as $u = v$, and $c_1 = n_{101} = |N_1(u) \cap N_0(v)| = |\{v\}| = 1$ as $v$ is adjacent to $u$. 

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2. If \( d(u, v) = i \) and \( w \) is adjacent to \( u \), then \( i - 1 \leq d(v, w) \leq i + 1 \). So
\[
\begin{align*}
\alpha_i + \beta_i + \gamma_i &= n_{1,1} + n_{1,i+1,i} + n_{1,i-1,i} = |N_1(u) \cap N_i(v)| + |N_1(u) \cap N_{i+1}(v)| + |N_1(u) \cap N_{i-1}(v)| = |N_1(u)| = d_1.
\end{align*}
\]

3. Suppose \( d(u, v) = i + 1 \), \( 1 \leq i \leq D - 1 \). Pick a path \( v, x, \ldots, u \) of length \( i + 1 \). Then \( d(x, u) = i \). If \( w \in N_{i-1}(x) \cap N_1(u) \), then \( w \in N_i(v) \cap N_1(u) \). So \( N_{i-1}(x) \cap N_1(u) \subseteq N_i(v) \cap N_1(u) \), that is, \( |N_{i-1}(x) \cap N_1(u)| \leq |N_i(v) \cap N_1(u)| \). In other words, \( c_i = n_{1,i-1,i} \leq n_{1,i,i+1} = c_{i+1} \) for \( 1 \leq i \leq D - 1 \).

4. Suppose \( d(u, v) = i \), \( 1 \leq i \leq D - 1 \). Pick a path \( v, x, \ldots, u \) of length \( i \). Then \( d(x, u) = i - 1 \). If \( w \in N_1(u) \cap N_{i+1}(v) \), then \( w \in N_i(v) \cap N_1(x) \). So \( N_1(u) \cap N_{i+1}(x) \subseteq N_1(u) \cap N_i(x) \), i.e. \( b_{i-1} = n_{1,i-1,i} = |N_1(u) \cap N_i(x)| \geq |N_1(u) \cap N_{i+1}(v)| = b_i \).

5. Pick any vertex \( v \). The number of edges from \( N_{i-1}(v) \) to \( N_i(v) \) is equal to the number of edges from \( N_i(v) \) to \( N_{i-1}(v) \), \( 1 \leq i \leq D \).

\[
\begin{align*}
N_{i-1}(v) & \quad N_i(v) \\
\cdot & \quad \cdot \\
v & \quad c_i \\
\cdot & \quad \cdot \\
b_{i-1} & \quad \cdot
\end{align*}
\]

The number of edges from \( N_i(v) \) to \( N_{i-1}(v) = c_i|N_i(v)| = c_i d_i \).

The number of edges from \( N_{i-1}(v) \) to \( N_i(v) = b_{i-1}|N_{i-1}(v)| = b_{i-1} d_{i-1} \).

So \( b_{i-1} d_{i-1} = c_i d_i \).

In the rest of this section, we will consider properties of Cayley graphs and transposition graphs.

**Theorem 4.10** Every Cayley graph is vertex-transitive.
Proof: Let $G(\Gamma, S)$ be a Cayley graph. Pick any two vertices $u$ and $v$ and define $\alpha : V \to V$ by $\alpha(x) = vu^{-1}x$, $x \in V$.

1. If $\alpha(x_1) = \alpha(x_2)$, then $vu^{-1}x_1 = vu^{-1}x_2$. So $x_1 = x_2$ and $\alpha$ is one-to-one.

2. For any $y \in V$, $\alpha(uv^{-1}y) = vu^{-1}uv^{-1}y = y$. Thus, $\alpha$ is onto.

3. If $x_1$ is adjacent to $x_2$, then $x_2 = x_1g$ for some $g \in S$. So $vu^{-1}x_2 = vu^{-1}x_1g$, or $\alpha(x_2) = \alpha(x_1)g$. That is, $\alpha(x_1)$ is adjacent to $\alpha(x_2)$ implying that $\alpha$ is an automorphism of $G(\Gamma, S)$.

Furthermore, $\alpha(u) = vu^{-1}u = v$. Hence, $\alpha$ is an automorphism that maps $u$ to $v$. \qed

Using the Cayley graph construction, we can obtain a vertex-transitive graph. If we use a group with certain properties, those properties may be reflected in the graph. The Proposition [2] is one such example.

**Proposition 4.11** Let $G$ be a connected graph. The subgroup $H$ of the automorphism group $\text{Aut}(G)$ acts regularly on $G$ if and only if $G$ is a Cayley graph $G(H, S)$ for some symbol set $S$ that generates $H$.

**Proof:** Suppose $G = G(H, S)$. For each $h \in H$, let $\alpha_h : H \to H$ be defined by $\alpha_h(x) = hx$. The mapping $\alpha_h$ is definitely a permutation. If $x$ is adjacent to $y$, then $y = xs$ for some $s \in S$. So $\alpha_h(y) = hy = hxs = \alpha_h(x)s$. That is, $\alpha_h(x)$ is adjacent to $\alpha_h(y)$. Therefore, $\alpha_h$ is an automorphism. The set of all $\alpha_h$ is a subgroup of $\text{Aut}(G)$ isomorphic to $H$. Let $\overline{H}$ be this subgroup. For any pair of vertices $x$ and $y$, $x, y \in H$. There is a unique $h \in H$ such that $hx = y$. Hence, there is an automorphism in $\overline{H}$, $\alpha_h$ such that $\alpha_h(x) = y$. Notice that if $e$ is the identity in $H$, then it is the unique element such that $\alpha_e(x) = x$. Hence, $\overline{H}$ acts on $G$ regularly.

Conversely, suppose $H$ is regular and $H \leq \text{Aut}(G)$. Let $V(G) = \{v_1, v_2, \ldots, v_n\}$ denote the vertex-set of $G$. Since $H$ is transitive, for each $i$, there exists $h_i \in H$ such that $h_i(v_1) = v_i$. Suppose $h_i(v_1) = h'_i(v_1) = v_i$. Then $h_i^{-1}h'_i(v_1) = h_i^{-1}h'_i(v_1) = v_i$. Since $H$ is regular, $h_i^{-1}h'_i = e$ and $h'_i = h_i$. This implies $h_i$ is the unique element in $H$ that maps $v_1$ to $v_i$. Now let $S = \{h_i \in H : v_i$ is adjacent to $v_1$ in $G\}$. Clearly $e \notin S$. If $h_i \in H$, then $v_i$ is
adjacent to $v_1$ and $h_i(v_i)$ is adjacent to $h_i(v_1) = v_i$. So $h_i^{-1}(h_i(v_i)) = h_i^{-1}(v_1)$ is adjacent to $h_i^{-1}(v_i) = v_i$. By the definition of $S$, $h_i^{-1} \in S$. Therefore, $S$ satisfies the conditions of being a symbol set. Let $\phi : G \to G(H, S)$ be defined by $\phi(v_i) = h_i$. Since there is a unique $h_i$ corresponding to $v_i$, $\phi$ is one-to-one. Since $H$ is transitive, $\phi$ is onto.

Suppose $v_i$ is adjacent to $v_j$. Then $h_i^{-1}(v_i) = v_1$ is adjacent to $h_i^{-1}(v_j) = h_i^{-1}h_j(v_1)$. So $h_i^{-1}h_j \in S$. Since $h_j = h_ih_i^{-1}h_j$, $h_i$ is adjacent to $h_j$ in $G(H, S)$. Conversely suppose $h_i$ is adjacent to $h_j$. Then $h_j = h_ih_i$, for some $h_i \in S$. Since $v_1$ is adjacent to $v_i$, $h_i(v_1) = v_i$ is adjacent to $h_i(v_1) = h_ih_i(v_1) = h_j(v_1) = v_j$. Thus, $G \cong G(H, S)$. □

Now we consider some relationships between Cayley graphs and transposition graphs [8].

**Lemma 4.12** Let $G(\Gamma_1, S_1)$ and $G(\Gamma_2, S_2)$ be two Cayley graphs on the permutation groups $\Gamma_1$ and $\Gamma_2$ acting on the sets $A_1$ and $A_2$, respectively. Let $\Gamma_1$ and $\Gamma_2$ be generated by the sets of transpositions $S_1$ and $S_2$ respectively, where $|S_1| = |S_2|$. If the transposition graphs $TG(A_1, S_1)$ and $TG(A_2, S_2)$ are isomorphic, then $G(\Gamma_1, S_1)$ and $G(\Gamma_2, S_2)$ are isomorphic too.

**Proof:** Let $\omega : TG(A_1, S_1) \to TG(A_2, S_2)$ be an isomorphism. Define $\beta : \Gamma_1 \to \Gamma_2$ by $\beta(u) = \omega \cdot u \cdot \omega^{-1}$. Then $\beta(u)$ is a composition of one-to-one and onto functions, so $\beta(u)$ is a permutation on $A_2$.

1. $\beta$ is one-to-one.

   If $\beta(u_1) = \beta(u_2)$, then $\omega \cdot u_1 \cdot \omega^{-1}(y) = \omega \cdot u_2 \cdot \omega^{-1}(y)$ for all $y \in A_2$, or $\omega(u_1(\omega^{-1}(y))) = \omega(u_2(\omega^{-1}(y)))$ for all $y \in A_2$. Since $\omega$ is one-to-one, $u_1(\omega^{-1}(y)) = u_2(\omega^{-1}(y))$ for all $y \in A_2$. Thus, $u_1(x) = u_2(x)$ for all $x \in A_1$, or $u_1 = u_2$.

2. $\beta$ is onto.

   Pick any $p \in \Gamma_2$. Let $u = \omega^{-1} \cdot p \cdot \omega$. Then $\beta(u) = \beta(\omega^{-1} \cdot p \cdot \omega) = \omega \cdot \omega^{-1} \cdot p \cdot \omega \cdot \omega^{-1} = p$.

3. $\beta$ preserves the adjacency.

   If $uv$ is an edge in $G(\Gamma_1, S_1)$, then $u = vs$ for some transposition $s = (i \ j) \in S_1$. Since $\omega$ is an isomorphism from $TG(A_1, S_1)$ to
Thus, \( \beta(u) = \omega \cdot u \cdot \omega^{-1} = \omega \cdot v \cdot s \cdot \omega^{-1} = \beta(v)(\omega(i)\omega(j)) \). That is, \( \beta(u) \) is adjacent to \( \beta(v) \).

Hence, \( \beta \) is an isomorphism from \( G(\Gamma_1, S_1) \) to \( G(\Gamma_2, S_2) \).

\[\text{Theorem 4.13} \quad \text{Let } G(\Gamma, S) \text{ be a Cayley graph on a permutation group } \Gamma \text{ acting on } A \text{ with the set of transpositions } S. \text{ If the transposition graph } TG(A, S) \text{ is edge-transitive, then } G(\Gamma, S) \text{ is } 1\text{-distance-transitive.}\]

Proof: Let \( e \) be the identity in \( \Gamma \) and let \( G = G(\Gamma, S) \). Let \( u, v, x, y \in \Gamma \) be such that \( uv \in E(G) \) and \( xy \in E(G) \). Since \( G \) is vertex-transitive, there exist automorphisms \( \alpha \) and \( \tau \) such that \( e = \alpha(u) \) and \( e = \tau(x) \). Let \( v' = \alpha(v) \) and \( y' = \tau(y) \). Since \( d(u, v) = d(e, v') = d(e, y') = d(x, y) = 1 \), \( v' \) and \( y' \) are transpositions.

Let \( v' = (i \ j) \) and \( y' = (l \ k) \). Since the transposition graph \( TG(A, S) \) is edge-transitive, there exists an automorphism \( \sigma \) such that \( l = \sigma(i) \) and \( k = \sigma(j) \), or \( k = \sigma(i) \) and \( l = \sigma(j) \). Notice that \( \sigma \) is a permutation on \( A \).

Define \( \beta : \Gamma \rightarrow \Gamma \) by \( \beta(p) = \sigma p \sigma^{-1} \). By the proof of Lemma 4.12, \( \beta \) is an automorphism of \( G \) so that \( \beta(v') = \sigma v' \sigma^{-1} = y' \) and \( \beta(e) = \sigma e \sigma^{-1} = e \).

Hence, \( \tau^{-1} \beta \alpha \) is an automorphism such that \( \tau^{-1} \beta \alpha(u) = x \) and \( \tau^{-1} \beta \alpha(v) = y \). That is, \( G(\Gamma, S) \) is \( 1 \)-distance-transitive.

\[\text{4.4 Set Graphs}\]

Although the construction of Cayley graphs gives us a way to build vertex-transitive graphs, it does not produce all vertex-transitive graphs. For example, the Petersen graph is not a Cayley graph. Before showing that the Petersen graph is a vertex-transitive graph but not a Cayley graph, let’s consider another construction of vertex-transitive graphs.

\[\text{Definition 4.14} \quad \text{Let } S = \{1, \ldots, n\}. \text{ The } \text{set graph} G(S, k) \text{ is a graph whose vertex-set is the set of all } k\text{-subsets of } S. \text{ Two vertices are adjacent if and only if the intersection of the corresponding subsets is empty.}\]
Example: Let $S = \{1, 2, 3, 4, 5\}$ and $k = 2$. We get the Petersen graph (Figure 4.1.)

Theorem 4.15 Every set graph is vertex-transitive.

Proof: Let $G(S, k)$ be a set graph. For any pair of vertices $u$ and $v$, we need to find an automorphism $\phi_{uv}$ such that $\phi_{uv}(u) = v$. Let $u = \{a_1, a_2, \ldots, a_k\}$ and $v = \{b_1, b_2, \ldots, b_k\}$. There exists a permutation $\sigma$ such that $\sigma(a_i) = b_i$ for all $1 \leq i \leq k$.

Let $\phi_{uv} : G \to G$ be defined by $\phi_{uv}(x) = \phi_{uv}\{x_1, x_2, \ldots, x_k\} = \{\sigma(x_1), \sigma(x_2), \ldots, \sigma(x_k)\}$. Suppose we have $\{\sigma(x_1), \sigma(x_2), \ldots, \sigma(x_k)\} = \{\sigma(y_1), \sigma(y_2), \ldots, \sigma(y_k)\}$, then we can relabel the elements so that $\sigma(x_i) = \sigma(y_i)$, $1 \leq i \leq k$. Since $\sigma$ is a permutation, $x_i = y_i$ for all $1 \leq i \leq k$, that is, $\{x_1, x_2, \ldots, x_k\} = \{y_1, y_2, \ldots, y_k\}$. Hence, $\phi_{uv}$ is one-to-one.

For any $\{z_1, z_2, \ldots, z_k\}$, Let $w = \{\sigma^{-1}(z_1), \sigma^{-1}(z_2), \ldots, \sigma^{-1}(z_k)\}$. Then $\phi_{uv}(w) = \{z_1, z_2, \ldots, z_k\}$. So $\phi_{uv}$ is onto.

If $\{x_1, x_2, \ldots, x_k\}$ is adjacent to $\{y_1, y_2, \ldots, y_k\}$, then $\{x_1, x_2, \ldots, x_k\} \cap \{y_1, y_2, \ldots, y_k\} = \emptyset$. Since $\sigma$ is a permutation, $\{\sigma(x_1), \sigma(x_2), \ldots, \sigma(x_k)\} \cap \{\sigma(y_1), \sigma(y_2), \ldots, \sigma(y_k)\} = \emptyset$. Therefore, $\phi_{uv}(\{x_1, x_2, \ldots, x_k\})$ is adjacent to $\phi_{uv}(\{y_1, y_2, \ldots, y_k\})$. 
Also, \( \phi_{uv}(\{a_1, a_2, \ldots, a_k\}) = \{\sigma(a_1), \sigma(a_2), \ldots, \sigma(a_k)\} = \{b_1, b_2, \ldots, b_k\} \).
Hence, \( \phi_{uv} \) is an automorphism mapping \( u \) to \( v \).

\[\]

Corollary 4.16 The Petersen graph is vertex-transitive.

Although Cayley graphs and set graphs are both vertex-transitive, in general, the Cayley graph construction cannot produce set graphs. Again we can show that the Petersen graph is not a Cayley graph.

Theorem 4.17 The Petersen graph is not a Cayley graph.

Proof: Suppose the Petersen graph is a Cayley graph \( G(\Gamma, S) \) for some group \( \Gamma \) and symbol set \( S \). From the fact of the group theory, there are only two possible groups of order \( 10[2, 13] \). Suppose \( \Gamma = \langle g \rangle \) for some element \( g \). Then \( \Gamma = \{e, g^1, g^2, \ldots g^9\} \). The symbol set can only be \( S_i = \{g^i, g^5, g^{-i}\} \), where \( i = 1, 2, 3, 4 \). Then \( e, eg^i, eg^5, eg^5g^{-i}, eg^5g^5g^{-i}g^5 = e \) is a 4-cycle. But the Petersen graph does not have any 4-cycle.

Then \( \Gamma = \{e, g, g^2, g^3, g^4, x, xg, xg^2, xg^3, xg^4\} \), where \( xg^ix = g^{-i} \) and \( x^2 = e \). Then the possible symbol sets are \( S_1 = \{xg^i, g, g^4\}, S_2 = \{xg^i, g^2, g^3\} \) and \( S_3 = \{xg^i, xg^k, xg^l\} \), where \( 0 \leq i, j, k, l \leq 4 \).

If the symbol set is \( S_1 \), then \( e, exg^i, exg^5g, exg^5g^ix, exg^5g^xg^5g = e \) is a 4-cycle. If the symbol set is \( S_2 \), then \( e, exg^i, exg^i g^2, exg^i g^2, exg^i g^2, exg^i g^2 xg^5g^2 = e \) is a 4-cycle. For \( S_3 \), we let \( S_3 = \{g_1, g_2, g_3\} \). We know that \( g_1^2 = g_2^2 = g_3^2 = e \). We can label the edges by the symbols in \( S_3 \) such that adjacent edges have different symbols assigned.

Consider the outermost 5-cycle. Two of the symbols in \( S_3 \) must be used twice. Without loss of generality, we label the outermost 5-cycle as Figure 4.2.

Then we can continue to label the edges, and finally we will get two adjacent edges having the same symbol. So we get the contradiction. Hence the Petersen graph is not a Cayley.

\[\]
Figure 4.2: Diagram for Theorem 4.17
Chapter 5

The Hypercube

The hypercube is usually considered to be an efficient networks for parallel computation. The construction of the hypercube is based on the binary numbers. As a consequence, routing algorithms for the hypercube are very easy to implement. Also, the hypercube is highly symmetric. In fact, it is distance-transitive. Furthermore, one can simulate most of the popular networks on the hypercube such as the grid and the binary tree. Hence, the hypercube is a good architecture for general purpose parallel systems [9].

5.1 Modelling

The hypercube is usually defined as follows [9].

Definition 5.1 Let $G(V, E)$ be a graph with $|V| = 2^r$ and $|E| = r2^{r-1}$ for some positive integer $r$. The vertices in $G$ are labelled with a binary sequence of length $r$. Two vertices are adjacent if and only if their binary sequences differ in precisely one bit. $G$ is called the $r$-dimensional hypercube and denoted as $Q_r$.

Figure 5.1 is a 3-dimensional hypercube. Besides the above definition, The hypercube can be defined as a Cayley graph too [8].

Proposition 5.2 Let $\Gamma = \langle (1 2), (3 4), \ldots, (2r-1 2r) \rangle$ be a subgroup of $S_{2r}$. Let $S = \{(1 2), (3 4), \ldots, (2r-1 2r)\}$ be the set of symbols. Then the Cayley graph $G(\Gamma, S) \cong Q_r$. 

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**Figure 5.1: The 3-dimensional Hypercube**

**Proof:** Let $v$ be a vertex in $Q_r$. Let $a_1a_2\cdots a_r$ be the binary sequence corresponding to $v$. Define $\phi: Q_r \rightarrow G(\Gamma, S)$ by $\phi(v) = \phi(a_1a_2\cdots a_r) = p_1p_2\cdots p_r$ where

$$p_i = \begin{cases} 
(2i-1 2i) & \text{if } a_i = 1 \\
\epsilon & \text{otherwise.} 
\end{cases}$$

Let $v_1, v_2 \in Q_r$ and $v_1 = a_1\cdots a_r$ and $v_2 = b_1\cdots b_r$. Since $(1 2), (3 4), \ldots, (2r-1 2r)$ are disjoint cycles, none of them can be generated by the others. Thus, if $v_1 \neq v_2$, then $a_1\cdots a_r \neq b_1\cdots b_r$. There are some $1 \leq i \leq r$ such that $a_i \neq b_i$. So $(2i-1 2i)$ is contained either in $\phi(v_1)$ or in $\phi(v_2)$ but not in both, that is, $\phi(v_1) \neq \phi(v_2)$.

Let $p \in \Gamma$. Let $a_1a_2\cdots a_r$ be a binary sequence such that

$$a_i = \begin{cases} 
1 & \text{if } (2i-1 2i) \text{ is in } p \\
0 & \text{otherwise.} 
\end{cases}$$

Then $\phi(a_1a_2\cdots a_r) = p$. Hence $\phi$ is a bijection.

Now if $a_1\cdots a_r$ is adjacent to $b_1\cdots b_r$, then there is exactly one $i$, $1 \leq i \leq r$ such that $a_j = b_j$ for $i \neq j$ and $a_i \neq b_i$. Then $\phi(a_1\cdots a_r) = \phi(b_1\cdots b_r)(2i - 1 2i)$. So $\phi(a_1\cdots a_r)$ is adjacent to $\phi(b_1\cdots b_r)$.  

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Figure 5.2: The 3-dimensional Hypercube Modelled as a Cayley Graph

Conversely, if $\phi(a_1 \cdots a_r)$ is adjacent to $\phi(b_1 \cdots b_r)$, then $\phi(a_1 \cdots a_r) = \phi(b_1 \cdots b_r)(2i - 1 2i)$ for some $i$. This implies that $a_j = b_j$ for $i \neq j$ and $a_i \neq b_i$. Hence, $a_1 \cdots a_r$ is adjacent to $b_1 \cdots b_r$. Therefore, $\phi$ is an isomorphism. Figure 5.2 is the Cayley graph version of the 3-dimensional hypercube.

One drawback of the hypercube is that the degree of each vertex is equal to $\log_2 |V|$. That means, when the network is getting bigger, the communication lines going out from the vertex will increase too. If a processor is designed for the $n$-dimensional hypercube, it cannot be used for the $8$-dimensional hypercube because four communication ports are missing from each processor. This drawback reduces the expandibility of the network.

5.2 Symmetry

However, the hypercube has very good symmetry properties. Since the hypercube is a Cayley graph, it is vertex-transitive. The transposition graph of the hypercube is a perfect matching, so it is 1-distance-transitive by Lemma 4.13. As mentioned before, the hypercube is in fact distance-transitive [2].
Lemma 5.3 The $r$-dimensional hypercube has diameter $r$.

Proof: For any pair of vertices $u$ and $v$ in $Q_r$, we can flip the necessary bits of $u$ one by one to get $v$. This also gives a route from $u$ to $v$. Thus, the diameter must be at most $r$. Since from $00\ldots0$ to $11\ldots1$ we have to flip at least $r$ bits, the diameter must be at least $r$. The result follows. 

Theorem 5.4 The hypercube is distance-transitive.

Proof: From the above Lemma, we know that the diameter of $Q_r$ is $r$. Since $Q_r$ is a Cayley graph, it is vertex-transitive. Let $u = p_1p_2 \cdots p_r$ be any vertex, where $p_j = (2j-1 2j)$ or $p_j = e$. Let $x$ and $y \in N_i(u)$, $0 \leq i \leq r$. There are precisely $i$ transpositions either in $x$ or in $u$ but not in both. Similarly, there are precisely $i$ transpositions either in $y$ or in $u$ but not in both.

Let $p_{k_1}, p_{k_2}, \ldots, p_{k_i}$ be the transpositions either in $x$ or in $u$ but not in both, and $p_{l_1}, p_{l_2}, \ldots, p_{l_i}$ be the transpositions either in $y$ or in $u$ but not in both. Let $p_{k_j} = (r_{k_j} r_{k_j}+1)$ and $p_{l_j} = (s_{l_j} s_{l_j}+1)$ for $1 \leq j \leq i$. Consider the mapping $\beta : ((1 2), \ldots (2r-1 2r)) \rightarrow ((1 2), \ldots (2r-1 2r))$ defined by

\[
\beta(v) = u(r_{k_1} s_{l_1})(r_{k_2} s_{l_2})(r_{k_3} s_{l_3})\cdots
\]

\[
(\begin{array}{c}
(r_{k_i} s_{l_i})(r_{k_i}+1 s_{l_i}+1)uv(r_{k_i} s_{l_i})(r_{k_i}+1 s_{l_i}+1)

(r_{k_2} s_{l_2})(r_{k_2}+1 s_{l_2}+1)\cdots(r_{k_i} s_{l_i})(r_{k_i}+1 s_{l_i}+1).
\end{array}
\]

Notice that every component in $\beta$ is the inverse of itself. We have

\[
\beta(u(t t+1)) = \begin{cases} 
   u(r_{k_j} r_{k_j}+1) & \text{if } t = s_{k_j} \\
   u(s_{l_j} s_{l_j}+1) & \text{if } t = r_{k_j} \\
   u(t t+1) & \text{otherwise} 
\end{cases}
\]

and

\[
\beta(u(t_1 t_1+1)(t_2 t_2+1)\cdots(t_m t_m+1))
\]

\[
= \beta(u(t_1 t_1+1))u\beta(u(t_1 t_1+1))\cdots u\beta(u(t_m t_m+1)).
\]

Also, $\beta^{-1} = \beta$. Clearly $\beta$ is a permutation on the vertices in $Q_r$. If $v_1$ is adjacent to $v_2$ in $Q_r$, then there is precisely one transposition in one of the $v_j$'s but not in both. There is also precisely one transposition in one of the $\beta(v_j)$'s but not in both. Hence, $\beta$ is an automorphism. Since $\beta(u) = u$ and $\beta(x) = y$, $\beta \in A_u$. By Lemma 4.5, $Q_r$ is distance-transitive. 

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5.3 Connectivity

To be a good network, a graph should have very high connectivity. The connectivity tells us how many nodes can be malfunctioned and the network is still connected. It also tells us how many node-disjoint paths between a pair of nodes. The more node-disjoint paths the network has, the more subproblems can be handled simultaneously.

**Definition 5.5** Let \( b \) be a binary digit. we define

\[
\bar{b} = \begin{cases} 
0 & \text{if } b = 1 \\
1 & \text{if } b = 0.
\end{cases}
\]

**Lemma 5.6** An \( n \)-regular graph has connectivity \( n \) if for any pair of vertices \( u \) and \( v \), there are \( n \) vertex-disjoint paths joining them.

**Proof:** If any two vertices are joined by \( n \) vertex-disjoint paths, then we should remove at least \( n \) vertices to disconnect the graph. However, \( G \) is \( n \)-regular, so \( G \) has connectivity \( n \).

Note: the converse of the lemma is also true (See the Menger's Theorem [3]), but we will not prove it here.

**Theorem 5.7** The \( r \)-dimensional hypercube \( Q_r \) has connectivity \( r \).

**Proof:** In the 2-dimensional hypercube, any pair of vertices are joined by two vertex-disjoint paths. We assume that any pair of vertices of \( Q_r \) are joined by \( r \) vertex-disjoint paths. Since \( Q_{r+1} \) is vertex-transitive, it is sufficient to show that there are \( r+1 \) vertex-disjoint paths joining the vertex \( 0 \cdots 0 \) and \( b_1 \cdots b_{r+1} \), where \( b_i \in \{0,1\}, 1 \leq i \leq r+1 \).

**Case 1:** The bit \( b_{r+1} = 0 \). Let \( S \) be the subgraph induced by the vertices whose \((r+1)\)th bit is 0, and let \( T \) be the subgraph induced by the vertices whose \((r+1)\)th bit is 1. \( S \) and \( T \) are \( r \)-dimensional hypercubes. There are \( r \) vertex-disjoint paths from \( 0 \cdots 0 \) to \( b_1 \cdots b_r 0 \) in \( S \). There is a path \( P \) from \( 0 \cdots 01 \) to \( b_1 \cdots b_r 1 \) in \( T \). Therefore, we have a path starting from \( 0 \cdots 0 \), passing through \( P \) and ending at \( b_1 \cdots b_r 0 \) in \( Q_{r+1} \). Together with the \( r \)
vertex-disjoint paths in $S$, we have $r + 1$ vertex-disjoint paths from $0 \cdots 0$ to $b_1 \cdots b_r 0$.

**Case 2:** The bit $b_{r+1} = 1$. We use the same definitions of $S$ and $T$. There are $r$ vertex-disjoint paths from $0 \cdots 0$ to $b_1 \cdots b_r 0$ in $S$. We remove $b_2 \cdots b_r 0$ from each of these paths. Let $P_i$ be the path from $0 \cdots 0$ to $b_{i-1} \overline{b_i} b_{i+1} \cdots b_r 0$, $1 \leq i \leq r$. We extend $P_i$ by adding the 2-path $b_1 \cdots b_{i-1} \overline{b_i} b_{i+1} \cdots b_r 0$, $b_1 \cdots b_{i-1} b_i b_{i+1} \cdots b_r 1$, $b_1 \cdots b_{i-1} b_i b_{i+1} \cdots b_r 1$ for $1 \leq i \leq r - 1$. We extend $P_r$ by adding the 2-path $b_1 \cdots b_{r-1} \overline{b_r} 0$, $b_1 \cdots b_{r-1} b_r 0$, $b_1 \cdots b_{r-1} b_r 1$. We translate $P_r$ to $T$ by changing the $(r + 1)$th bit of each vertex to 1 and call it $P'_r$. Then we have the path from $0 \cdots 0$ passing through $P'_r$ to $b_1 \cdots b_{r-1} b_r 1$. Hence, we have $r + 1$ vertex-disjoint paths from $0 \cdots 0$ to $b_1 \cdots b_r 1$. By Lemma 5.6, the result follows.

**Corollary 5.8** The edge-connectivity of the $r$-dimensional hypercube is $r$.

**Proof:** Suppose the edge-connectivity is $k$, where $k < r$. Let $T$ be the set of $k$ edges whose removal will disconnect the graph. Then for each edge in $T$ we can remove one of the incident vertices to disconnect the graph. But it is a contradiction. Since the $r$-dimensional hypercube is $r$-regular. The edge-connectivity must be $r$.

**5.4 Other Known Results**

It is not difficult to see that the hypercube is bipartite. We can get the bipartition by letting one of the partition sets be the set of vertices with an even number of 1's.

**Definition 5.9** Let $G = (X, Y)$ be a bipartite graph. If for any pair of vertices, $x \in X$ and $y \in Y$, there is a Hamilton path from $x$ to $y$, then $G$ is said to be Hamilton-laceable.

**Definition 5.10** Let $G$ be a graph. If for any pair of vertices $x$ and $y$ in $G$, there is a Hamilton path from $x$ to $y$, then $G$ is said to be Hamilton-connected.
**Theorem 5.11** The r-dimensional hypercube is Hamilton-laceable.

*Proof:* $Q_2$ is Hamilton-laceable. Assume $Q_r$ is Hamilton-laceable. Let $u = u_1 u_2 \ldots u_{r+1}$ and $v_1 v_2 \ldots v_{r+1}$ be any vertices in $Q_{r+1}$, where $u_i, v_i \in \{0,1\}$ for $1 \leq i \leq r$. Suppose $u_i \neq v_i$. Let $P_1$ be a Hamilton path from $u_1 \ldots u_{i-1} u_{i+1} \ldots u_{r+1}$ to $u_1 \ldots u_{i-1} \bar{u}_{i+1} \ldots u_{r+1}$ in $Q_r$ and $P_2$ a be a Hamilton path from $u_1 \ldots u_{i-1} \bar{u}_{i+1} \ldots u_{r+1}$ to $v_1 \ldots v_{i-1} v_{i+1} \ldots v_{r+1}$ in another copy of $Q_r$ (These exist because $u_1 \ldots u_{i-1} \bar{u}_{i+1} \ldots u_{r+1}$ and $v_1 \ldots v_{i-1} v_{i+1} \ldots v_{r+1}$ differ in an odd number of bits.)

Now we insert $u_i$ in the $i$th position for every vertex in $P_1$ and denote the new path as $P_1(u_i)$. We also insert $v_i$ in the $i$th position for every vertex in $P_2$ and denote the new path as $P_2(v_i)$. Clearly $P_1(u_i)P_2(v_i)$ is a Hamilton path from $u_1 \ldots u_{r+1}$ to $v_1 \ldots v_{r+1}$ in $Q_{r+1}$. By induction, the result follows. \(\square\)

**Corollary 5.12** The r-dimensional hypercube is hamiltonian.
Chapter 6

The Butterfly Network

The butterfly network is one of the modifications of the hypercube. It inherits some of the properties of the hypercube, but its degree is bounded. In fact, the butterfly graph is a 4-regular graph. Like the hypercube, the butterfly network can simulate most of the networks with bounded degree with acceptable slowdown [9].

6.1 Modelling

The following definition of the butterfly graph is taken from [9].

Definition 6.1 Let $G(V, E)$ be a graph with $|V| = r2^r$ and $|E| = r2^{r+1}$ for some positive integer $r$. The vertices in $G$ are labelled as $(w, i)$, where $w$ is a binary sequence of length $r$ that is called the row of the vertex, and $i$ is the level of the vertex ($1 \leq i \leq r$). Two vertices $(w, i)$ and $(w', i')$ are adjacent if and only if either

1. $w = w'$ and $i' \equiv i \pm 1 \pmod{r}$ or

2. $w$ and $w'$ differ in precisely the $i'$th bit when $i' \equiv i + 1 \pmod{r}$ or $w$ and $w'$ differ in precisely the $i$th bit when $i' \equiv i - 1 \pmod{r}$

$G$ is called the $r$-dimensional butterfly graph and denoted as $B_r$.  

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Figure 6.1: The 3-dimensional Butterfly Graph
Figure 6.1 exhibits the 3-dimensional butterfly graph. Notice that if we identify the vertices in the same row, and remove all the loops and multiple edges, we will get an $r$-dimensional hypercube.

**Proposition 6.2** Let $B_r$ be an $r$-dimensional butterfly graph. If $G$ is the graph obtained from $B_r$ by identifying the vertices in the same row and removing all the loops and multiple edges, then $G \cong Q_r$.

**Proof:** This follows directly from the definitions of the butterfly graph and the hypercube. 

Again, the butterfly graph can be modelled as a Cayley graph. But the group structure will not be as simple as the one for the hypercube. For the butterfly graph, each vertex has two coordinates. The first one is related to the one in the hypercube. Thus, we will extend the group for the hypercube to the one for the butterfly graph $[11, 1]$.

Let $\Gamma_r = \{(p, i) : p = p_1 \cdots p_r, p_i \in \{(2j-1 2j) : j = 1, \ldots, r\} \cup \{e\}, 0 \leq i \leq r - 1\}$ and $S = \{(1 2), (3 4), \ldots, (2r-1 2r)\}$. Define $\pi_i : S \rightarrow S$ by $\pi_i((2j-1 2j)) = (2(i+j)-1 2(i+j))$ reduced modulo $2r$ and $\pi_i(e) = e$. Clearly, $\pi_{i+k}(p) = \pi_i(\pi_k(p))$.

Define a binary operator $\cdot$ as follows:

$$(p, i) \cdot (p', i') = (p_1 \cdots p_r, i) \cdot (p'_1 \cdots p'_r, i') = (p_1 \cdots p_r \pi_i(p'_1) \cdots \pi_i(p'_r), i + i'),$$

where $i + i'$ is reduced modulo $r$.

**Proposition 6.3** We have that $(\Gamma_r, \cdot)$ is a group.

**Proof:** Note that

$$
\begin{align*}
[(p, i) \cdot (p', i')] \cdot (p'', i'') &= [(p_1 \cdots p_r, i) \cdot (p'_1 \cdots p'_r, i')] \cdot (p'', i'') \\
&= (p_1 \cdots p_r \pi_i(p'_1) \cdots \pi_i(p'_r), i + i') \cdot (p'', i'') \\
&= (p_1 \cdots p_r \pi_i(p'_1) \cdots \pi_i(p'_r), i + i') \cdot (p''_1 \cdots p''_r, i'') \\
&= (p_1 \cdots p_r \pi_i(p'_1) \cdots \pi_i(p'_r), i + i' + i'') \\
&= (p_1 \cdots p_r, i) \cdot [(p'_1 \cdots p'_r, i') \cdot (p''_1 \cdots p''_r, i'')] \\
&= (p, i) \cdot [(p', i') \cdot (p'', i'')].
\end{align*}
$$
So · is associative.

Since \((e, 0) \cdot (p, i) = (p, i) \cdot (e, 0) = (p, i)\), \((e, 0)\) is the identity.

For any \((p, i) = (p_1 \cdots p_r, i)\),

\[
(p_1 \cdots p_r, i) \cdot (\pi_{-i}(p_1) \cdots \pi_{-i}(p_r), -i) = (p_1 p_1 \cdots p_r p_r, 0)
\]

\[
= (e, 0),
\]

and every element has an inverse. \(\square\)

Now we can model the butterfly graph as a Cayley graph.

**Proposition 6.4** Let \(\Gamma_r\) be the group in Proposition 6.3 and \(S = \{(e, 1), (e, r-1), ((1 2), 1), ((2r-1 2r), r-1)\}\). The Cayley Graph \(G(\Gamma_r, S) \cong B_r\).

**Proof:** Let \((w, i)\) be a vertex in \(B_r\), where \(w = a_1 a_2 \cdots a_r\) is a binary sequence of length \(r\). Define \(\phi : B_r \to G(\Gamma, S)\) by \(\phi((w, i)) = \phi((a_1 \cdots a_r, i)) = (p_1 \cdots p_r, i)\), where

\[
p_i = \begin{cases} 
2i - 2i & \text{if } a_i = 1 \\
e & \text{otherwise.}
\end{cases}
\]

Let \((w_1, i_1), (w_2, i_2) \in B_r\) where \(w_1 = a_1 \cdots a_r\) and \(w_2 = b_1 \cdots b_r\). If \(i_1 \neq i_2\), then clearly \(\phi((w_1, i_1)) \neq \phi((w_2, i_2))\). If \(w_1 \neq w_2\), then using the same argument in Proposition 5.2, We have \(\phi((w_1, i_1)) \neq \phi((w_2, i_2))\).

Let \((p, i) \in \Gamma\). Let \(a_1 \cdots a_r\) be a binary sequence such that

\[
a_i = \begin{cases} 
1 & \text{if } (2i-1 2i) \text{ is in } p \\
0 & \text{otherwise.}
\end{cases}
\]

Then \(\phi((a_1 \cdots a_r, i)) = (p, i)\). Hence \(\phi\) is a bijection.

If \((w_1, i_1)\) is adjacent to \((w_2, i_2)\), then there are four cases.

**Case 1:** \(w_1 = w_2\) and \(i_1 \equiv i_2 + 1 \pmod{r}\). Then \(\phi((w_1, i_1))(e, 1) = \phi((w_2, i_2))\). Thus, \(\phi((w_1, i_1))\) is adjacent to \(\phi((w_2, i_2))\).

**Case 2:** \(w_1\) and \(w_2\) differ in the \(i_1\)th bit and \(i_1 \equiv i_2 + 1 \pmod{r}\). Then \(\phi((w_1, i_1))((1 2), 1) = \phi((w_2, i_2))\). Thus, \(\phi((w_1, i_1))\) is adjacent to \(\phi((w_2, i_2))\).

Conversely, if \(\phi((w_1, i_1))\) is adjacent to \(\phi((w_2, i_2))\), there are four cases.
Case 1: if $\phi(\langle w_1, i_1 \rangle)(e, 1) = \phi(\langle w_2, i_2 \rangle)$, then $w_1 = w_2$ and $i_2 \equiv i_1 + 1 \pmod{r}$. Hence, $\langle w_1, i_1 \rangle$ is adjacent to $\langle w_2, i_2 \rangle$.

Case 2: if $\phi(\langle w_1, i_1 \rangle)(e, r - 1) = \phi(\langle w_2, i_2 \rangle)$, then $w_1 = w_2$ and $i_2 \equiv i_1 + r - 1 \pmod{r}$, or $i_1 \equiv i_2 + 1 \pmod{r}$. Hence, $\langle w_1, i_1 \rangle$ is adjacent to $\langle w_2, i_2 \rangle$.

Case 3: if $\phi(\langle w_1, i_1 \rangle)((1, 2), 1) = \phi(\langle w_2, i_2 \rangle)$, then $w_1$ and $w_2$ differ in the $i_1$th bit and $i_2 \equiv i_1 + 1 \pmod{r}$. Hence, $\langle w_1, i_1 \rangle$ is adjacent to $\langle w_2, i_2 \rangle$.

Case 4: if $\phi(\langle w_1, i_1 \rangle)((2r - 1, 2r), r - 1) = \phi(\langle w_2, i_2 \rangle)$, then $w_1$ and $w_2$ differ in the $(i_1 - 1)$th bit and $i_2 \equiv i_1 + r - 1 \pmod{r}$. That is, $w_1$ and $w_2$ differ in the $i_2$th bit and $i_1 \equiv i_2 + 1 \pmod{r}$. Hence, $\langle w_1, i_1 \rangle$ is adjacent to $\langle w_2, i_2 \rangle$.

\[\square\]

Corollary 6.5 All butterfly graphs are vertex-transitive.

Proof: Since all Cayley graph are vertex-transitive, the result follows. \[\square\]

6.2 Symmetry

Although the butterfly graph is derived from the hypercube, unfortunately it does not inherit all the symmetry properties from the hypercube. In fact, the butterfly graph is not even edge-transitive. This means that it is not distance-transitive or $k$-distance-transitive because those transitivities imply edge-transitivity.

Theorem 6.6 Butterfly graphs are not edge-transitive for $r \geq 3$.

Proof: For $r \geq 3$, consider the $r$-cycle,

$$\langle 00 \cdots 0, 1 \rangle \langle 00 \cdots 0, 2 \rangle \cdots \langle 00 \cdots 0, r - 1 \rangle \langle 00 \cdots 0, r \rangle \langle 00 \cdots 0, 1 \rangle.$$
Each edge \((00 \cdots 0, i)(00 \cdots 0, i + 1)\) in this cycle lies in the unique 4-cycle,
\[(00 \cdots 0, i) \langle 00 \cdots 01 \cdots 0, i + 1 \rangle \langle 00 \cdots 01 \cdots 0, i \rangle \langle 00 \cdots 0, i + 1 \rangle \langle 00 \cdots 0, i \rangle.
\]
These 4-cycles are edge-disjoint. If the butterfly graph is edge-transitive, the edge \((00 \cdots 0, 1)(10 \cdots 0, r)\) must lie in an \(r\)-cycle with the same property described above. Suppose such a cycle exists. Then \((00 \cdots 0, 1)\) and \((10 \cdots 0, r)\) are the first and the second vertex.
The third vertex cannot be $(10 \cdots 0, 1)$ because $(00 \cdots 0, 1)(10 \cdots 0, r)$ and $(10 \cdots 0, r)(10 \cdots 0, 1)$ are in the same 4-cycle

$$(00 \cdots 0, 1)(10 \cdots 0, r)(10 \cdots 0, 1)(00 \cdots 0, r)(00 \cdots 0, 1).$$

Hence, the third vertex must be either $(10 \cdots 0, r - 1)$ or $(10 \cdots 01, r - 1)$. The fourth vertex cannot be in the $r$th level. Otherwise, the second and the third edge will be in the same 4-cycle (see the figure above). Similarly, the fifth vertex cannot be in the $(r - 1)$th level. Otherwise, the third and the fourth edge will be in the same 4-cycle, and so on. It forces the last vertex $v_r$ to be in the second level. Since the path from $(00 \cdots 0, 1)$ to the last vertex $v_r$ passes through the $r$th level exactly once at the second vertex $(10 \cdots 0, r)$, the first bit of the last vertex is 1. Hence, $v_r$ is not adjacent to $(00 \cdots 0, 1)$. That is, the cycle in fact does not exist. Therefore, the butterfly graph is not edge-transitive.

\[\square\]

### 6.3 Topological Structure

Besides the properties of symmetry, topological properties are also very important in studying interconnection networks. For example, people are unlikely to use a network with the large diameter because in general it takes longer time to communicate. Furthermore, it may be good news for anyone who wants to pipeline the job if the network is hamiltonian. We will now consider the topological properties of the butterfly graph.

**Proposition 6.7** The $r$-dimensional butterfly graph has girth 4, for $r \geq 4$.

**Proof:** Since $(00 \cdots 0, 1)(010 \cdots 0, 2)(010 \cdots 0, 1)(00 \cdots 0, 2)(00 \cdots 0, 1)$ is a 4-cycle, the butterfly graph has girth at most 4. Now suppose there is a triangle in the butterfly graph. Let $(w_1, i_1)$, $(w_2, i_2)$ and $(w_3, i_3)$ be the vertices of this triangle. Then $|i_1 - i_2| = 1$, $|i_2 - i_3| = 1$ and $|i_1 - i_3| = 1$ which is impossible unless $r = 3$.

\[\square\]

Before determining the diameter of the butterfly graph, we present a simple routing algorithm that is completely based on the definition of the butterfly graph.

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Let $s = \langle w, i \rangle$ be the source node and $t = \langle w', i' \rangle$ be the destination. Let $w = a_1 a_2 \cdots a_r$ and $w' = b_1 b_2 \cdots b_r$. If $q = c_1 c_2 \cdots c_r$, then denote $\text{flip}(q, k) = c_1 c_2 \cdots c_{k-1} \bar{c}_k c_{k+1} \cdots c_r$, where

$$
\bar{c}_k = \begin{cases} 
0 & \text{if } c_k = 1 \\
1 & \text{if } c_k = 0.
\end{cases}
$$

Algorithm 6.1: Simple Routing Algorithm for Butterfly Graph

The level indices are reduced modulo $r$ and in the range from 1 to $r$.

1. Let $p \leftarrow i$, $l \leftarrow 0$ and $q_0 \leftarrow w$.
2. $l \leftarrow l + 1$, $p \leftarrow p + 1$.
3. If $a_p \neq b_p$, then

   $q_l \leftarrow \text{flip}(q_{l-1}, p),$

   else

   $q_l \leftarrow q_{l-1}$.

4. If $p \neq i$ then go to step 2.

5. If $(i' - i) < (i - i')$ then

   the route is:

   $$\langle w, i \rangle \langle q_1, i + 1 \rangle \langle q_2, i + 2 \rangle \cdots \langle q_r, i \rangle \langle q_r, i + 1 \rangle \cdots \langle q_r, i' \rangle,$$

   else

   the route is:

   $$\langle w, i \rangle \langle q_1, i + 1 \rangle \langle q_2, i + 2 \rangle \cdots \langle q_r, i \rangle \langle q_r, i - 1 \rangle \cdots \langle q_r, i' \rangle.$$

Example: In the 4-dimensional butterfly graph, the route from $\langle 1011, 2 \rangle$ to $\langle 0110, 4 \rangle$ will be

$$\langle 1011, 2 \rangle \langle 1011, 3 \rangle \langle 1010, 4 \rangle \langle 0010, 1 \rangle \langle 0110, 2 \rangle \langle 0110, 1 \rangle \langle 0110, 4 \rangle.$$
The above algorithm changes the bits in \( w \) one by one to get \( w' \). Then it continues the path in the row \( w' \) to get the correct level \( i' \). The algorithm actually works. Notice that the loop from step 2 to step 4 runs \( r \) times. In step 5, it augments the path with at most \( \lceil \frac{r}{2} \rceil \) vertices. Thus the length of the route that the algorithm produces is at most \( r + \lceil \frac{r}{2} \rceil = \lfloor \frac{3r}{2} \rfloor \). With this result, we can show that the diameter of the butterfly graph is \( \lfloor \frac{3r}{2} \rfloor \).

**Theorem 6.8** The \( r \)-dimensional butterfly graph has diameter \( \lfloor \frac{3r}{2} \rfloor \).

*Proof:* By Algorithm 6.1, we know that the length of the path between any two vertices is at most \( \lfloor \frac{3r}{2} \rfloor \), so the diameter of the butterfly graph is at most \( \lfloor \frac{3r}{2} \rfloor \) too. Now consider the shortest path from \( (00 \cdots 0, r) \) to \( (11 \cdots 1, 1) \). We must change all 0’s of the source to 1’s and move from level \( r \) to level \( \lfloor \frac{r}{2} \rfloor \). No matter how we move, we have to take at least \( r + \lceil \frac{r}{2} \rceil = \lfloor \frac{3r}{2} \rfloor \) steps. Thus, the diameter of the butterfly is at least \( \lfloor \frac{3r}{2} \rfloor \). The result follows. \( \Box \)

Algorithm 6.1 is not so good because it always produces a walk of length at least \( r \). Of course, we can identify the repeated vertices and remove the vertices in between to make the walk be a path to get some improvement. However, the result is still not a shortest path in general. For example, consider a path between \( (000, 3) \) and \( (011, 1) \) in the 3-dimensional butterfly graph. Figure 6.2 shows that Algorithm 6.1 does not give the shortest path.

Before presenting a shortest path algorithm, we consider a simple optimization problem. Given an \( n \)-cycle

\[ v_0e_1v_1e_2v_3 \cdots v_{n-2}e_{n-2}v_{n-1}e_nv_0, \]

where \( v_0, v_1, \ldots, v_{n-1} \) are vertices and \( e_1, e_2, \ldots, e_n \) are edges, let \( A \subseteq E \) be a subset of the edge-set, and \( s, t \in V \) be any two vertices. The problem is to find a shortest walk from \( s \) to \( t \) so that it covers all the edges in \( A \) (See Figure 6.3).

**Lemma 6.9** The shortest walk covers each vertex at most twice.

*Proof:* Suppose the walk covers the vertex \( v \) three times. Then the walk will look like one of the diagrams in Figure 6.4.
Figure 6.2: The Routes from (000, 3) to (011, 1)
Figure 6.3: Diagram for the Shortest Walk Problem

Figure 6.4: Diagram for Lemma 6.9
Clearly, the walks from $a$ to $b$ are redundant in all cases. If we remove all the redundant parts, we can get a shorter walk that covers $v$ at most once. If the walk covers the vertex more than three times, we can use this operation repeatedly to reduce the number of times the walk covers the vertex.

The walk has to start from $s$ and stop at $t$, so it must contain a path from $s$ to $t$. Based on Lemma 6.9, the shortest walk will look like one of the diagrams in Figure 6.5.

Suppose $A = \{e_1, e_2, \ldots, e_k\}$. Then the walk we need to consider will be $sP_1e_1P_2tP_3e_2P_3$ as illustrated in Figure 6.6.

Note that $P_1$ and $P_2$ may be empty and $P_1$ and $P_2$ are the reverse paths of $P_1$ and $P_2$, respectively. There are in fact $k$ walks to check, and the shortest one will be the shortest walk from $s$ to $t$ that covers all edges in $A$. 

\[ \square \]
Now we can model the shortest path problem in the butterfly graph as the above optimization problem. Given any two vertices \((w, i)\) and \((w', i')\) where \(w = a_1a_2 \ldots a_r\) and \(w' = b_1b_2 \ldots b_r\), let

\[
C = v_r e_1 v_1 e_2 v_2 e_3 \cdots e_{r-2} v_{r-2} e_{r-1} v_{r-1} e_r v_r
\]

be an \(r\)-cycle, \(A = \{e_i : a_i \neq b_i\}\), \(s = v_i\) and \(t = v_{i'}\).

Once we get the shortest walk in \(C\), we can transform the solution to the shortest path of the butterfly graph. Let

\[
W = v_{j_0} e_{i_1} v_{j_1} e_{i_2} v_{j_2} e_{i_3} \cdots e_{i_p} v_{j_p},
\]

where \(j_0 = i\) and \(j_p = i'\), be a shortest walk that covers all edges in \(A\)

**Algorithm 6.2**: The shortest path algorithm for the butterfly graph.

The level indices are reduced modulo \(r\) and in the range from 1 to \(r\).

1. Let \(W = v_{j_0} e_{i_1} v_{j_1} e_{i_2} v_{j_2} e_{i_3} \cdots e_{i_p} v_{j_p}\), where \(j_0 = i\) and \(j_p = i'\), be a shortest walk that covers all edges in \(A\)
2. Let $q_0 \leftarrow w$.

3. For $l = 1$ to $p$
   if $e_{il} \in A$ then
   
   $q_l \leftarrow flip(q_{l-1}, i_l)$, $A \leftarrow A - \{e_{il}\}$,
   else
   
   $q_l \leftarrow q_{l-1}$.

4. The shortest path from $(w, i)$ to $(w', i')$ is

   $\langle q_0, j_0 \rangle \langle q_1, j_1 \rangle \cdots \langle q_p, j_p \rangle$.

Example: Consider the same example mentioned before. We want to find the shortest path from $(000, 3)$ to $(011, 1)$ in a 3-dimensional butterfly graph. The associated 3-cycle is:

Clearly, the shortest walk is $v_3e_3v_2e_2v_1$. Using algorithm 6.2, we get

$\langle 000, 3 \rangle \langle 001, 2 \rangle \langle 011, 1 \rangle$

which is the desired result.

The above algorithm works because from $(w, i)$ to $(w', i')$ we have to change $w$ bit by bit to get $w'$. Each time we change one bit, $i$ will be changed to be either $i + 1$ or $i - 1$. Finally we have to make $i$ become $i'$ too. Consider the following diagram:
We want to start at row $i$ and finish at row $i'$. If we move to the right at row $j$, we can change the $(j + 1)$th bit ($e_{j+1}$ is marked) or keep it the same ($e_{j+1}$ is not marked). If we move to the left at row $j$, we can change the $j$th bit ($e_j$ is marked) or keep it the same ($e_j$ is not marked). It is exactly the shortest walk problem that we have discussed.

Of course, Algorithm 6.1 is much easier to implement. It is not necessary to pre-determine the route before the node sends the message. This also means that no extra memory is required to store the route if Algorithm 6.1 is used. Thus, there is some trade-off between Algorithm 6.1 and Algorithm 6.2.

6.4 Hamilton Cycles and Hamilton Paths

Next we show that the butterfly graph is hamiltonian and discuss an algorithm to determine a Hamilton cycle. The following is an algorithm to determine a Hamilton cycle in butterfly graphs.

**Algorithm 6.3**: Hamilton cycle algorithm for butterfly graphs

The level indices additions are reduced modulo $r$ and in the range from 1 to $r$.

1. Set $(w_0, i_0) \leftarrow (00 \cdots 0, r)$.

2. For $l = 1$ to $r^2$
   
   let $k = \begin{cases} 
   0 & \text{if } w_{l-1} = 00 \cdots 0 \\
   \max\{j : j\text{th bit of } w_{l-1} \text{ is 1}\} & \text{otherwise.}
   \end{cases}$

   If $i_{l-1} < k$ then
3. The Hamilton cycle is

\[ \langle w_0, i_0 \rangle \langle w_1, i_1 \rangle \cdots \langle w_{r-2}, i_{r-2} \rangle. \]

Figure 6.7 is a Hamilton cycle of the 5-dimensional butterfly graph generated by Algorithm 6.3.

**Theorem 6.10** Algorithm 6.3 generates a Hamilton cycle of the butterfly graph and hence, all butterfly graphs are hamiltonian.

**Proof:** We need to show that Algorithm 6.3 generates all vertices in the butterfly graph. Suppose we start at the vertex \( \langle a_1 \cdots a_{r-1}0, r \rangle \). Algorithm 6.3 will generate the following:

\[ \langle a_1 \cdots a_{r-1}1, 1 \rangle \langle a_1 \cdots a_{r-1}1, r - 1 \rangle \langle a_1 \cdots a_{r-1}1, r - 2 \rangle \cdots \langle a_1 \cdots a_{r-1}1, 1, r \rangle \langle a_1 \cdots a_{r-1}0, r - 1 \rangle. \]

So all vertices in row \( a_1 \cdots a_r \) are generated. The path also contains the vertex, \( \langle a_1 \cdots a_{r-1}0, r \rangle \) and returns to \( \langle a_1 \cdots a_{r-1}0, r - 1 \rangle \). Assume that if we start at the vertex \( \langle a_1 \cdots a_j0 \cdots 0, r \rangle \), then Algorithm 6.3 will generate the path that contains all vertices in row \( a_1 \cdots a_jb_{j+1} \cdots b_r \), \( b_l \in \{0,1\} \) and \( \langle a_1 \cdots a_j0 \cdots 0, l \rangle \), where \( j + 1 \leq l \leq r \). Also the path will return to \( \langle a_1 \cdots a_j0 \cdots 0, j \rangle \).

Now suppose we start at \( \langle a_1 \cdots a_{j-1}0 \cdots 0, r \rangle \). Then by the assumption, Algorithm 6.3 will generate the path that contains all vertices in row \( a_1 \cdots a_{j-1}0b_{j+1} \cdots b_r \), \( b_l \in \{0,1\} \) and \( \langle a_1 \cdots a_{j-1}0 \cdots 0, l \rangle \), where \( j + 1 \leq l \leq r \). The path will return to \( \langle a_1 \cdots a_{j-1}0 \cdots 0, j \rangle \). The path will continue to \( \langle a_1 \cdots a_{j-1}10 \cdots 0, j - 1 \rangle \langle a_1 \cdots a_{j-1}10 \cdots 0, j - 2 \rangle \cdots \langle a_1 \cdots a_{j-1}10 \cdots 0, r \rangle \). Then by the assumption again, Algorithm 6.3 will generate the path that contains all vertices in row \( a_1 \cdots a_{j-1}1b_{j+1} \cdots b_r \), \( b_l \in \{0,1\} \) and \( \langle a_1 \cdots a_{j-1}10 \cdots 0, l \rangle \), where \( j + 1 \leq l \leq r \). The path will then return to \( \langle a_1 \cdots a_{j-1}10 \cdots 0, j \rangle \) and then it continues to \( \langle a_1 \cdots a_{j-1}0 \cdots 0, j - 1 \rangle \). By induction, if we start at \( \langle 00 \cdots 0, r \rangle \), Algorithm 6.3 will generate all vertices in the butterfly graph. \( \square \)
Figure 6.7: A Hamilton Cycle in the 5-dimensional Butterfly Graph
The butterfly graph is not only hamiltonian, but also hamilton-connected for odd dimension and hamilton-laceable for even dimension. We are going to show this stronger result [12].

**Theorem 6.11** The butterfly graph $B_r$ is Hamilton-laceable when $n$ is even.

**Proof**: First we show that if $B_{r-2}$ is Hamilton-laceable, then $B_r$ is also Hamilton-laceable. Let $R_{r-2}(x,y)$ be the rows $xyb_3 \cdots b_r$, where $b_i \in \{0,1\}$ for all $3 \leq i \leq r$. Since $B_r$ is vertex-transitive, it is sufficient to show that there is Hamilton path from $(0 \cdots 0, r)$ to any vertex $v = (a_1 \cdots a_r, l)$, where $l$ is odd.

**Case 1**: The bits $a_1a_2 = 00$. If $l \geq 3$, then let $P_{r-2}$ be a Hamilton path from $(0 \cdots 0, r - 2)$ to $(a_3 \cdots a_r, l - 2)$ in $B_{r-2}$. If $l < 3$, then let $P_{r-2}$ be a Hamilton path from $(0 \cdots 0, r - 2)$ to $(a_3 \cdots a_r, 1)$ such that the edges $(a_3 \cdots a_r, r - 2) \langle \bar{a}_3 \cdots a_r, 1 \rangle$ and $(a_3 \cdots a_r, r - 2) \langle a_3 \cdots a_r, 1 \rangle$ are not in $P_{r-2}$.

First we construct a path in $R_{r-2}(x,y)$ from $P_{r-2}$. We relabel each vertex $(bl \cdots br-2, i)$ in $P_{r-2}$ to $(xybl \cdots br-2, i+2)$ and augment level 1 and level 2 to $P_{r-2}$. We replace the edge $(xyb_1 \cdots b_{r-2}, i + 2)$ by the path

$$
(xy\bar{b}_1 \cdots b_{r-2}, 3)(xyb_1 \cdots b_{r-2}, 2)(xyb_1 \cdots b_{r-2}, 1)(xyb_1 \cdots b_{r-2}, r),
$$

and replace the edge $(xyb_1 \cdots b_{r-2}, r)(xyb_1 \cdots b_{r-2}, 3)$ by the path

$$
(xyb_1 \cdots b_{r-2}, 3)(xyb_1 \cdots b_{r-2}, 2)(xyb_1 \cdots b_{r-2}, 1)(xyb_1 \cdots b_{r-2}, r).
$$

If $l = 1$, we also put a path

$$
(xyb_1 \cdots b_{r-2}, 3)(xyb_1 \cdots b_{r-2}, 2)(xyb_1 \cdots b_{r-2}, 1)
$$

(it is possible because of the choice of $P_{r-2}$). Let the resultant path be $S_{r-2}(0,0)$.

Using Algorithm 6.3, we can get a path from $(xy0 \cdots 0, r)$ to $(xy0 \cdots 0, 2)$ which contains all but $(xy0 \cdots 0, 1)$ in $R_{r-2}(x,y)$. Therefore, by putting the edge $(xy0 \cdots 0, 1)(xy0 \cdots 0, 2)$, we get a Hamilton path from $(xy0 \cdots 0, r)$ to $(xy0 \cdots 0, 1)$ in $R_{r-2}(x,y)$. We denote this path by $T_{r-2}(0,0)$.

Since $P_{r-2}$ is a Hamilton path, it must use an edge whose endpoints are in the first and the $r$th level. That means $S_{r-2}$ contains an edge whose
Figure 6.8: Diagram 1 for Theorem 6.11
endpoints are in the first and the second level. Let $f$ be that edge. Now we connect $S_{r-2}$ and three $T_{r-2}$’s together to get $P_r$ as Figure 6.8 shown.

For the isolated vertices $(00d_1 \ldots d_{r-2}, 1)$ and $(00d_1 \ldots d_{r-2}, 2)$. We remove the edge $(01d_1 \ldots d_{r-2}, 1)(01d_1 \ldots d_{r-2}, 2)$ and add a path

$$(01d_1 \ldots d_{r-2}, 1)(00d_1 \ldots d_{r-2}, 2)(00d_1 \ldots d_{r-2}, 1)(01d_1 \ldots d_{r-2}, 2).$$

Notice that $P_r$ also satisfies the restriction of $P_{r-2}$.

**Case 2:** The bits $a_1a_2 \neq 00$ and $a_3a_4 \ldots a_r \neq 0 \ldots 0$. We first connect the paths as Figure 6.9 shown. Then we use the same method as in case 1 to join the isolated vertices into the path. Again the resultant path satisfies the restriction of $P_{r-2}$.
Case 3: The bits $a_1a_2 \neq 00$ and $a_3 \cdots a_r = 0 \cdots 0$. Let $\overline{S}_{r-2}(x, y)$ be the path from $(xy1 \cdots 1, r)$ to the vertices in row $xy0 \cdots 0$ in $R_{r-2}(x, y)$ obtained by reversing the rows of $S_{r-2}(x, y)$ which starts from $(xy0 \cdots 0, r)$ to the vertices in row $xy1 \cdots 1$. We define $\overline{T}_{r-2}(x, y)$ in a similar manner. We also let $S_{0,r-2}(x, y)$ be the path from $(xy0 \cdots 0, r)$ to $(xy1 \cdots 1, 1)$. Now we connect the paths as Figure 6.10 and join the isolated vertices to get $P_r$.

![Figure 6.10: Diagram 3 for Theorem 6.11](image)

Notice that all of $P_r$ satisfies the restriction of $P_{r-2}$. When $r = 2$, we have the following Hamilton paths.
By induction, the result follows.

\[\Box\]

**Theorem 6.12** The Butterfly graph $B_r$ is Hamilton-connected when $r$ is odd.

*Proof:* we will use the notation in Theorem 6.11. The proof is basically the same as Theorem 6.11. Suppose $B_{r-2}$ is Hamilton-connected. We want to find a Hamilton path from $(0\ldots0,r)$ to any vertex $(a_1\ldots a_r, l)$. The same procedure in Theorem 6.11 will be used to get the Hamilton path in $B_r$. There is one exception: $l = 2$. When $l = 2$ and $a_1\ldots a_r \neq 0\ldots0$. We let $P_{r-2}$ be the path from $(0\ldots0,r-2)$ to $(a_1\ldots a_r, r-2)$ such that it does not have the edges $(a_1\ldots a_r,1)(a_1\ldots a_r,r-2)$ and $(a_1\ldots a_r,r-2)(a_1\ldots a_r,1)$. Then we construct $S_{r-2}(x,y)$ as before and add a path $(xya_1\ldots a_r, r)(xya_1\ldots a_r, 1)(xya_1\ldots a_r, 2)$. The rest will be the same as Theorem 6.11.

If $l = 2$ and $a_1\ldots a_r = 0\ldots0$, the we let $S_{r-2}(x,y)$ be the path from $(xy0\ldots0, r)$ to $(xy1\ldots1, r)$. We also let $S_{0,r-2}(x,y)$ be the path from $(xy0\ldots0, r)$ to $(xy1\ldots1, 1)$. Then we connect the paths as Figure 6.11 and Figure 6.12 shown.

Notice that all $P_r$ satisfy the restriction of $P_{r-2}$. For the Hamilton paths starting from $(000, 3)$ in $B_3$. We can use the following $S_1(x,y)$ to generate them.

By induction, the result follows.

\[\Box\]

**6.5 Connectivity**

Unlike the hypercube, the butterfly graph has the bounded connectivity because of its fixed degree.

**Theorem 6.13** The connectivity of $r$-dimensional butterfly graph is 4.
Figure 6.11: Diagram 1 for Theorem 6.12
Figure 6.12: Diagram 2 for Theorem 6.12
Proof: Let \( u, v \) be any two vertices in \( B_r \). Since \( B_r \) is vertex-transitive, we can assume that \( u = (0 \ldots 0, r) \). Using the Algorithm 6.3, we can get a Hamilton cycle \( (0 \ldots 0, r) AvB(0 \ldots 0, r) \). If \( A \) or \( B \) is empty, then \( B_r - \{(0 \ldots 0, r), v\} \) is connected. We assume that both \( A \) and \( B \) are non-empty. Then \( (0 \ldots 01, r - 1) \in A \) and \( (10 \ldots 0, 1) \in B \). Consider the paths

\[ P_1 : (0 \ldots 01, r - 1)(0 \ldots 01, r)(10 \ldots 01, 1)(10 \ldots 01, 2) \ldots (10 \ldots 01, r - 1) \]
\[ (10 \ldots 0, r)(10 \ldots 0, 1). \]
\[ P_2 : (0 \ldots 01, r - 2)(0 \ldots 011, r - 2) \ldots (01 \ldots 11, 1)(11 \ldots 1, r) \]
\[ (11 \ldots 100, r - 1)(11 \ldots 100, r - 2)(11 \ldots 1000, r - 3) \ldots (10 \ldots 0, 1). \]
\[ P_3 : (0 \ldots 01, r - 2)(0 \ldots 01, r - 3) \ldots (0 \ldots 01, 1)(10 \ldots 01, r) \]
\[ (10 \ldots 0, r - 1)(10 \ldots 0, r - 2) \ldots (10 \ldots 0, 1). \]

Those are the vertex-disjoint paths from \( (0 \ldots 01, r - 1) \) to \( (10 \ldots 0, 1) \). There must exist an edge from \( A \) to \( B \). That is, \( B_r - \{(0 \ldots 0, r), v\} \) is connected. Therefore, the connectivity of \( B_r \) is at least 3.

Let \( u, v \) and \( w \) be any three vertices in \( B_r \). Again, we can assume that \( u = (0 \ldots 0, r) \). Using Algorithm 6.3 and relabelling \( v \) and \( w \) if necessary, we can get a Hamilton cycle \( (0 \ldots 0, r) AvBwC(0 \ldots 0, r) \). If any two of \( A, B \) and \( C \) are empty, then \( B_r - \{(0 \ldots 0, r), v, w\} \) is still connected. If only \( B \) is empty, then we can use the same argument above to show that the graph is still connected. Suppose only \( A \) is empty. Then \( B \) contains \( (0 \ldots 01, r - 2) \) and \( C \) contains \( (10 \ldots 0, 1) \). Consider the paths

\[ P_1 : (0 \ldots 01, r - 2)(0 \ldots 011, r - 1)(10 \ldots 011, 1)(10 \ldots 011, 2) \ldots \]
\[ (10 \ldots 011, r - 2)(10 \ldots 01, r - 1)(10 \ldots 0, r)(10 \ldots 0, 1). \]
\[ P_2 : (0 \ldots 01, r - 2)(0 \ldots 0101, r - 3) \ldots (01 \ldots 101, 1)(11 \ldots 101, r) \]
\[ (11 \ldots 100, r - 1)(11 \ldots 100, r - 2)(11 \ldots 1000, r - 3) \ldots (10 \ldots 0, 1). \]
\[ P_3 : (0 \ldots 01, r - 2)(0 \ldots 01, r - 3) \ldots (0 \ldots 01, 1)(10 \ldots 01, r) \]
\[ (10 \ldots 0, r - 1)(10 \ldots 0, r - 2) \ldots (10 \ldots 0, 1). \]

Those are the vertex-disjoint paths from \( (0 \ldots 01, r - 1) \) to \( (10 \ldots 0, 1) \). There must exist an edge from \( B \) to \( C \). That is, \( B_r - \{(0 \ldots 0, r), v, w\} \) is connected. Therefore, the connectivity of \( B_r \) is at least 3.

Suppose only \( C \) is empty. Then \( A \) contains \( (0 \ldots 01, r - 1) \) and \( B \) contains \( (110 \ldots 0, 2) \). Consider the paths

\[ P_1 : (0 \ldots 01, r - 1)(0 \ldots 01, r)(10 \ldots 01, 1)(110 \ldots 01, 2) \ldots \]
\[ (110 \ldots 01, r - 1)(110 \ldots 0, r)(110 \ldots 0, r - 1) \ldots (110 \ldots 0, 2). \]
\[ P_2 : (0 \ldots 01, r - 1)(0 \ldots 011, r - 2) \ldots (01 \ldots 11, 1)(11 \ldots 1, r) \]
\[ (11 \ldots 10, r - 1) \ldots (110 \ldots 0, 2). \]
\[ P_3 : (0 \ldots 01, r - 1)(0 \ldots 01, r - 2) \ldots (0 \ldots 01, 1)(10 \ldots 01, r) \]
\[ (10 \ldots 0, r - 1)(10 \ldots 0, r - 2) \ldots (10 \ldots 0, 2)(110 \ldots 0, 1)(110 \ldots 0, 2). \]
Those are the vertex-disjoint paths from \((0 \cdots 01, r - 1)\) to \((110 \cdots 0, 2)\). There must exist an edge from \(A\) to \(B\). That is, \(B_r - \{(0 \cdots 0, r), v, w\}\) is connected.

Finally, suppose all \(A, B\) and \(C\) are non-empty. Then \((0 \cdots 01, r - 1)\) \(\in\) \(A\) and \(C\) contains \((10 \cdots 0, 1)\). We can use the three vertex-disjoint path from \((0 \cdots 01, r - 1)\) to \((10 \cdots 0, 1)\) described above to show that there is a path from \(A\) to \(C\). If there is an edge from \(A\) or \(C\) to \(B\), then \(B_r - \{(0 \cdots 0, r), v, w\}\) is connected. Otherwise, since the connectivity of \(B_r\) is at least 3, either \((0 \cdots 0, r - 1)\) or \((0 \cdots 0, 1)\) is in \(B\). For the first case, we have the following three vertex-disjoint paths from \((0 \cdots 01, r - 1)\) to \((0 \cdots 0, r - 1)\):

\[
\begin{align*}
P_1: & \quad (0 \cdots 01, r - 1)(0 \cdots 01, r)(0 \cdots 0, r - 1). \\
P_2: & \quad (0 \cdots 01, r - 1)(0 \cdots 01, r - 2)(0 \cdots 011, r - 1)(0 \cdots 010, r) \\
& \quad (0 \cdots 010, 1)(0 \cdots 010, r - 2)(0 \cdots 0, r - 1). \\
P_3: & \quad (0 \cdots 01, r - 1)(0 \cdots 01, r - 2)(0 \cdots 011, r - 1)(0 \cdots 01, r) \\
& \quad (10 \cdots 01, r - 1)(10 \cdots 0, r)(0 \cdots 0, 1)(0 \cdots 0, 2)(0 \cdots 0, r - 1).
\end{align*}
\]

For the second case, we have the following three vertex-disjoint paths from \((0 \cdots 01, r - 1)\) to \((0 \cdots 0, 1)\):

\[
\begin{align*}
P_1: & \quad (0 \cdots 01, r - 1)(0 \cdots 01, r)(0 \cdots 0, r - 1)(0 \cdots 0, 1). \\
P_2: & \quad (0 \cdots 01, r - 1)(0 \cdots 01, r - 2)(0 \cdots 01, 1)(0 \cdots 01, r) \\
& \quad (10 \cdots 01, r - 1)(10 \cdots 0, r)(0 \cdots 0, 1). \\
P_3: & \quad (0 \cdots 01, r - 1)(0 \cdots 011, r - 2)(0 \cdots 011, 1)(0 \cdots 011, 2) \\
& \quad (010 \cdots 1, 3)(010 \cdots 0, r)(010 \cdots 0, 1)(010 \cdots 0, 2)(0 \cdots 0, 1).
\end{align*}
\]

Hence, there must be a path from \(A\) to \(B\). \(B_r - \{(0 \cdots 0, r), v, w\}\) is connected. Since \(B_r\) is 4-regular, it has connectivity 4.

\[\square\]

**Corollary 6.14** The edge-connectivity of \(r\)-dimensional butterfly graph is 4.

**Proof:** Since \(B_r\) is 4-regular and has connectivity 4, the result follows.  \[\square\]
Chapter 7

Cube-connected-cycles

In the previous chapter, we discuss an extension of the hypercube, the butterfly graph. It not only inherits some properties from the hypercube, but also has bounded degree. In this chapter, we will discuss another extension of the hypercube which also has bounded degree. Consider the following operation:

Let $v$ be a vertex of a hypercube $Q_r$, $r \geq 3$. We replace each vertex by an $r$-cycle as illustrated above. Then the resultant graph will become a 3-regular graph and is called the $r$-dimensional cube-connected-cycles. Figure 7.1 is a 3-dimensional cube-connected-cycles.

7.1 Modelling

The following is the formal definition of $r$-dimensional cube-connected-cycles [9].
**Definition 7.1** Let $G(V, E)$ be a graph with $|V| = r2^r$ and $|E| = 3r2^{r-1}$ for some positive integer $r$. The vertices in $G$ are labelled by $(w, i)$, where $w$ is a binary sequence of length $r$ that denotes the row of the vertex and $i$ is the level of the vertex ($1 \leq i \leq r$). Two vertices $(w, i)$ and $(w', i')$ are adjacent if and only if either:

1. $w = w'$ and $i' \equiv i \pm 1 \pmod{r}$ or
2. $w$ and $w'$ differ in precisely the $i$th bit and $i' = i$.

The graph is called *r-dimensional cube-connected-cycles* and is denoted as $CCC_r$.

It is not surprising that cube-connected-cycles are also Cayley graphs. By comparing the cube-connected-cycles and the butterfly graphs, we can discover some similarities between these two classes of graphs. In fact, the group used to generate the cube-connected cycles is exactly the one for the butterfly graphs [11, 1].

![Figure 7.1: The 3-dimensional cube-connected-cycles](image-url)
Proposition 7.2 Let $\Gamma_r$ be the group in Proposition 6.3 and $S = \{(e, 1), (e, r-1), ((1 2), 0)\}$. The Cayley graph $G(\Gamma_r, S) \cong CCC_r$.

Proof: Let $(w, i)$ be a vertex in $CCC_r$, where $w = a_1a_2 \cdots a_r$ is a binary sequence of length $r$. Define $\phi : CCC_r \to G(\Gamma, S)$ by $\phi((w, i)) = \phi(\langle a_1 \cdots a_r, i\rangle) = (p_1 \cdots p_r, i)$, where

$$p_i = \begin{cases} (2i-1 \ 2i) & \text{if } a_i = 1 \\ e & \text{otherwise.} \end{cases}$$

We have already shown that this mapping is a bijection in Proposition 6.4. Let $(w_1, i_1), (w_2, i_2) \in CCC_r$, where $w_1 = a_1 \cdots a_r$ and $w_2 = b_1 \cdots b_r$. If $(w_1, i_1)$ is adjacent to $(w_2, i_2)$, then there are two cases.

Case 1: $w_1 = w_2$ and $i_1 \equiv i_2 + 1 \pmod{r}$. Then $\phi((w_1, i_1))(e, 1) = \phi((w_2, i_2))$. Therefore, $\phi((w_1, i_1))$ is adjacent to $\phi((w_2, i_2))$.

Case 2: $w_1$ and $w_2$ differ in the $i_1$th bit and $i_1 = i_2$. Then $\phi((w_1, i_1))(1 2), 0) = \phi(w_2, i_2)$. Thus, $\phi((w_1, i_1))$ is adjacent to $\phi((w_2, i_2))$.

Conversely, if $\phi((w_1, i_1))$ is adjacent to $\phi((w_2, i_2))$, there are three cases.

Case 1: If $\phi((w_1, i_1))(e, 1) = \phi((w_2, i_2))$, then $w_1 = w_2$ and $i_1 \equiv i_2 + 1 \pmod{r}$. Hence, $(w_1, i_1)$ is adjacent to $(w_2, i_2)$.

Case 2: If $\phi((w_1, i_1))(e, r-1) = \phi((w_2, i_2))$, then $w_1 = w_2$ and $i_2 \equiv i_1 + r - 1 \pmod{r}$, or $i_1 \equiv i_2 + 1 \pmod{r}$. Hence, $(w_1, i_1)$ is adjacent to $(w_2, i_2)$.

Case 3: If $\phi((w_1, i_1))(1 2), 0) = \phi((w_2, i_2))$, then $w_1$ and $w_2$ differ in the $i_1$th bit and $i_2 = i_1$. Hence, $(w_1, i_1)$ is adjacent to $(w_2, i_2)$.

This shows that $\phi$ is an isomorphism.

Corollary 7.3 All cube-connected-cycles are vertex-transitive.

7.2 Symmetry

As with the butterfly graph, the cube-connected-cycles is not edge-transitive. This also means that the cube-connected-cycles cannot be distance-transitive or even $k$-distance-transitive.
Theorem 7.4 The CCC, is not edge-transitive for $r \geq 3$.

Proof: Consider the $r$-cycle of row $(00\ldots 0)$ when $r \neq 8$ (figure 7.2.)

$$(00\ldots 0,1)(00\ldots 0,2)\cdots (00\ldots 0,r-1)(00\ldots 0,r)(00\ldots 0,1).$$

Each edge $(00\ldots 0,i)(00\ldots 0,i+1)$ in this cycles lies in the unique 8-cycle

$$(0\ldots 0,i)(0\ldots 010\ldots 0,i)(0\ldots 010\ldots 0,i+1)(0\ldots 0110\ldots 0,i+1)
\tag*{\text{(i-1)}}$$

$$(0\ldots 010\ldots 0,i)(0\ldots 010\ldots 0,i+1)(0\ldots 010\ldots 0,i+1)
\tag*{\text{(i-1)}}$$

If the cube-connected-cycles is edge-transitive, then we can find an automorphism that maps the edge $(0\ldots 0,1)(0\ldots 0,2)$ to the edge $(0\ldots 0,r-2)(0\ldots 0100,r-2)$. Thus, the edge $(0\ldots 0,r-2)(0\ldots 0100,r-2)$ must lie in an $r$-cycle with the same property. Let $W$ be this $r$-cycle. The edge $(0\ldots 0,r-2)(0\ldots 100,r-2)$ lies in two 8-cycles (see figure 7.2). If the edge $g$ is in $W$, then

$$(0\ldots 0100,1)(0\ldots 0100,2)\cdots (0\ldots 0100,r)$$

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will be the 8-cycle containing $g$. Similarly, if the edge $h$ is in $W$, then the same cycle
\[(0 \cdots 0100, 1)(0 \cdots 0100, 2) \cdots (0 \cdots 0100, r)\]
will be the 8-cycle containing $h$. Hence, $r$ must be 8, but it is a contradiction.

When $r = 8$, there are two types of 8-cycles. We call the 8-cycle of each row be the 8-cycle of type I, and the 8-cycle lying in more than one row be the 8-cycle of type II. Two 8-cycles are said to be adjacent if they have an edge in common. Consider those 8-cycles of type II in Figure 7.2 again. Any two disjoint 8-cycles of type II have only one common adjacent 8-cycle of type I which is the cycle of row $00 \cdots 0$. If the cube-connected-cycles is edge-transitive, again we want to find an automorphism that maps the edge $(0 \cdots 0, 1)(0 \cdots 0, 2)$ to the edge $(0 \cdots 0, r - 2)(0 \cdots 0100, r - 2)$. In this case, we can map the cycle $(0 \cdots 0, 1)(0 \cdots 0, 2) \cdots (0 \cdots 0, r)(0 \cdots 0, 1)$ to one of the cycles $R$ or $S$ (see figure 7.3). If $R$ is the image, then the disjoint cycles $S$ and $T$ have two common adjacent cycles $R$ and $V$. This means that this case is impossible. If $S$ is the image, then the disjoint cycles $D$ and $E$ have two common adjacent cycles $U$ and $S$. Again it is impossible. Hence, it is impossible to map the edge $(00 \cdots 0, 1)(00 \cdots 0, 2)$ to the edge

Figure 7.3: Diagram 2 for Theorem 7.4
\langle 00 \cdots 0, r-2 \rangle \langle 00 \cdots 0100, r-2 \rangle$. That is, none of the cube-connected-cycles are edge-transitive.

### 7.3 Topological Structure

We now consider topological properties of the cube-connected-cycles.

**Proposition 7.5** The $r$-dimensional cube-connected cycle has girth $r$ for $r \leq 8$ and girth 8 for $r > 8$.

**Proof:** Consider a cycle $\langle w_1i_1 \rangle \langle w_2i_2 \rangle \cdots \langle w_ki_k \rangle \langle w_1i_1 \rangle$. If all $w_i$'s are the same, we will get a $r$-cycle. Since any two rows are joined by at most one edge, it is impossible that a cycle lies in precisely two rows. Similarly, any three rows are joined by at most two edges, so it is impossible that a cycle is lies in precisely three rows. Suppose the cycle lies in precisely four rows. The cycle must use at least one edge from each row. This implies the cycle must have length at least 8.

The edge $\langle a_1 \cdots a_r, i \rangle \langle a_1 \cdots a_r, i+1 \rangle$ lies in the $r$-cycle

$$\langle a_1 \cdots a_r, 0 \rangle \cdots \langle a_1 \cdots a_r, r \rangle \langle a_1 \cdots a_r, 0 \rangle$$

and an 8-cycle

$$\langle a_1 \cdots a_{i+1} \cdots a_r, i \rangle \langle a_1 \cdots a_{i+1} \cdots a_r, i+1 \rangle \langle a_1 \cdots a_{i+1} \cdots a_r, i \rangle \langle a_1 \cdots a_{i+1} \cdots a_r, i+1 \rangle$$

$$\langle a_1 \cdots \overline{a}_{i+1} \cdots a_r, i \rangle \langle a_1 \cdots \overline{a}_{i+1} \cdots a_r, i \rangle \langle a_1 \cdots a_{i+1} \cdots a_r, i \rangle \langle a_1 \cdots a_{i+1} \cdots a_r, i \rangle$$

Hence, the result follows. \qed

As the butterfly graphs and the cube-connected-cycles are generated by the same group, there are some similarities between these two classes of graphs. For instance, if we modify the simple routing algorithm for the butterfly graph (Algorithm 6.1), we will get the following version for the cube-connected-cycles.

Let $s = \langle w, i \rangle$ be the source node and $t = \langle w', i' \rangle$ be the destination, where $w = a_1a_2 \cdots a_r$, and $w' = b_1b_2 \cdots b_r$.

**Algorithm 7.1:** Simple Routing Algorithm for Cube-Connected-Cycles
The level indices are reduced modulo $r$ and in the range from 1 to $r$.

1. Let $p \leftarrow i$, $l \leftarrow 0$, $q_0 \leftarrow w$ and $i_0 \leftarrow i$.

2. If $|i + 1 - i'| < |i - 1 - i'|$ then
   
   $s = 1,$

   else

   $s = -1.$

3. $l \leftarrow l + 1.$

4. If $a_p \neq b_p$, then
   
   \begin{align*}
   q_l & \leftarrow \text{flip}(q_{l-1}, p), \\
   i_l & \leftarrow i_{l-1},
   \end{align*}

   else

   \begin{align*}
   q_l & \leftarrow q_{l-1}, \\
   i_l & \leftarrow i_{l-1} + s, \\
   p & \leftarrow p + s.
   \end{align*}

5. If $p \neq i - s$, then go to step 2.

6. If $a_{i-s} \neq b_{i-s}$ then
   
   \begin{align*}
   l & \leftarrow l + 1, \\
   q_l & \leftarrow \text{flip}(q_{l-1}, i - s), \\
   i_l & \leftarrow i_{l-1}.
   \end{align*}

7. The route is

   \[(w, i)(q_1, i_1)(q_2, i_2) \cdots (q_l, i_l)(q_l, i_l - s)(q_l, i_l - 2s) \cdots (q_l, i').\]

The idea of this algorithm is the same as that of Algorithm 6.1. It changes the bits in $w$ one-by-one to get $w'$. Then it continues the route in row $w'$ to get the correct level $i'$. Notice that the loop from step 3 to step 5 in Algorithm 7.1 runs at most $2r - 1$ times. Because of step 2, the algorithm augments the route with at most $\left\lceil \frac{r-2}{2} \right\rceil$ vertices if $r \geq 4$, or 1 if $r = 3$. Hence,
the algorithm always gives a route of length at most $2r - 1 + \left\lfloor \frac{r-2}{2} \right\rfloor = \left\lfloor \frac{5r-4}{2} \right\rfloor$, when $r \geq 4$, and at most $2(3) - 1 + 1 = 6$, when $r = 3$. This leads to the following result [7].

**Theorem 7.6** The CCC has diameter $\left\lfloor \frac{5r-4}{2} \right\rfloor$, when $r \geq 4$, and 6 when $r = 3$.

**Proof:** When $r = 3$, the length of the shortest route from $(000, 3)$ to $(111, 2)$ is 6. The result follows. When $r \geq 4$, the algorithm gives a route of length at most $\left\lfloor \frac{5r-4}{2} \right\rfloor$. Consider the route from $(00\cdots 0, r)$ to $(11\cdots 1, \left\lfloor \frac{r}{2} \right\rfloor)$. The route must hit at least $r$ rows. The route also comes across each level at least once. The $i$th level is only connected to the $(i - 1)$th and $(i + 1)$th level. This forces the route to traverse at least $r - 1 + \left\lfloor \frac{r}{2} \right\rfloor - 1 = \left\lfloor \frac{3r-4}{2} \right\rfloor$ vertices if the route starts from level $r$, ends at level $\left\lfloor \frac{r}{2} \right\rfloor$ and crosses every level at least once. Hence, the length of the route from $(00\cdots 0, r)$ to $(11\cdots 1, \left\lfloor \frac{r}{2} \right\rfloor)$ is at least $r + \left\lfloor \frac{3r-4}{2} \right\rfloor = \left\lfloor \frac{5r-4}{2} \right\rfloor$. The result follows. 

Algorithm 7.1 does not give the shortest route in general. For instance, consider the route from $(0000, 4)$ to $(1100, 2)$. Algorithm 7.1 will generate the following route:

$$(0000, 4)(0000, 3)(0000, 2)(0100, 2)(0100, 1)(1100, 1)(1100, 2).$$

However, the shortest route is

$$(0000, 4)(0000, 1)(1000, 1)(1000, 2)(1100, 2).$$

We are now going to investigate the shortest path algorithm for cube-connected-cycles. In fact, the idea is exactly the same as that for the butterfly graphs. We know that if the source and the destination differ in $k$ bits, the route must traverse at least $k$ rows. There is no way to reduce this number. However, we can minimize the number of levels that the route hits. First consider a simple optimization problem. Given an $n$-cycle

$$C = v_1v_2\cdots v_r,$$
where $v_1v_2\cdots v_r$ are vertices, let $A \subseteq V$ be a subset of the vertex-set, and $s, t \in V$ be any two vertices. The problem is to find a shortest walk from $s$ to $t$ so that it covers all the vertices in $A$ (see figure 7.4).

This optimization problem is almost the same as that for the butterfly graphs. In fact, we can use the same method to solve this problem. This solution corresponds to the minimum number of levels that the route must hit.

Now we can model the shortest path problem in the cube-connected-cycles as the above optimization problem. Given any two vertices $(w, i)$ and $(w', i')$, where $w = a_1a_2\cdots a_r$ and $w' = b_1b_2\cdots b_r$, let

$$C = v_1v_2v_3\cdots v_r$$

be an $r$-cycle, $A = \{v_k : a_k \neq b_k, 1 \leq k \leq r\}$, $s = v_i$ and $t = v_{i'}$.

Suppose we have a shortest walk in $C$ that covers all vertices in $A$. Let

$$W = v_{j_0}v_{j_1}v_{j_2}v_{j_3}\cdots v_{j_p},$$

where $j_0 = i$ and $j_p = i'$, be such a shortest walk. We can now transform the solution to the shortest path in the cube-connected-cycles.

**Algorithm 7.2:** The shortest path algorithm for the cube-connected-cycles.

The level indices are reduced modulo $r$ and in the range from 1 to $r$. 

---

Figure 7.4: Diagram for the Shortest Walk Problem.
1. Let $W = v_{j_0} v_{j_1} v_{j_2} v_{j_3} \cdots v_{j_p}$ be the shortest walk of the associated optimization problem.

2. Let $q_0 \leftarrow w$, $l \leftarrow 0$, $k \leftarrow 0$ and $i_0 \leftarrow i$.

3. $l \leftarrow l + 1$.

4. If $v_{j_k} \in A$ then
   
   \[ q_l \leftarrow \text{flip}(q_{l-1}, j_k), \; i_l \leftarrow j_k, \; A \leftarrow A - \{v_{j_k}\}, \]
   
   else
   
   \[ k \leftarrow k + 1, \; q_l \leftarrow q_{l-1}, \; i_l \leftarrow j_k. \]

5. If $k < p$ then go to step 2.

6. If $v_{j_p} \in A$ then
   
   \[ l \leftarrow l + 1, \]
   \[ q_l \leftarrow \text{flip}(q_{l-1}, j_p), \; i_l \leftarrow j_p. \]

7. The shortest path is
   
   \[ \langle q_0, i_0 \rangle \langle q_1, i_1 \rangle \langle q_2, i_2 \rangle \cdots \langle q_l, i_l \rangle. \]

Example: Consider the same example mentioned before. We are looking for a shortest path from $(0000, 4)$ to $(1100, 2)$. The associated 4-cycle is shown below.

Clearly the shortest walk is $W = v_4 v_1 v_2$. Algorithm 7.2 will generate the path

\[ (0000, 4) \langle 0000, 1 \rangle (1000, 1) \langle 1000, 2 \rangle (1100, 2). \]

which is the desired result.

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7.4 Hamilton Cycles

The cube-connected-cycles is again hamiltonian. The result has been proven by R. Stong [11].

Theorem 7.7 All cube-connected-cycles are hamiltonian.

Proof: Let \( R_j \) be the subgraph of an \( r \)-dimensional cube-connected-cycles induced by the rows

\[
0 \cdots 0a_{j+1}a_{j+2} \cdots a_r, \quad a_i \in \{0, 1\}, \quad j + 1 \leq i \leq r.
\]

Suppose \( R_j \) is hamiltonian. Consider \( R_{j-2} \). It consists of four copies of \( R_j \) which are \( R_j(0, 0) \), \( R_j(0, 1) \), \( R_j(1, 0) \) and \( R_j(1, 1) \).

![Graph Diagram]

The vertices \( \langle 0 \cdots 0 xy0 \cdots 0, j - 1 \rangle \) and \( \langle 0 \cdots 0 xy0 \cdots 0, j \rangle \) have degree 2 in \( R_j(x, y) \). Therefore, we can let \( P_j(x, y) \) be a Hamilton path starting at \( \langle 0 \cdots 0 xy0 \cdots 0, j - 1 \rangle \) and ending at \( \langle 0 \cdots 0 xy0 \cdots 0, j \rangle \). Let \( \overline{P}_j(x, y) \) be the reverse of this Hamilton path. Then we can get a Hamilton cycle of \( R_{j-2} \), namely,

\[
P_j(0, 0)\overline{P}_j(0, 1)P_j(1, 1)\overline{P}_j(1, 0).
\]

When \( r \) is even, \( R_r \) is an \( r \)-cycle, so the result follows. When \( r \) is odd, we have the Hamilton cycle of \( R_{r-3} \) (Figure 7.5). By induction, the result follows. \( \square \)
Figure 7.6 illustrates Hamilton cycles of the 4-dimensional and the 5-dimensional cube-connected-cycles.

Theorem 7.7 gives a recursive construction of a Hamilton cycle of the cube-connected-cycles. In practice, we may want to have an algorithm that can determine the next vertex of the Hamilton cycle directly. With the algorithm, we do not need to store a Hamilton cycle in memory.

In the recursive construction, the even cases and the odd cases are separated. In the following algorithm, we still separate it into two different cases. For the even cases, let \( v = \langle a_1 a_2 \cdots a_{2p}, i \rangle \). We choose

\[
\langle 0 \cdots 0, r \rangle \langle 0 \cdots 0, r - 1 \rangle \cdots \langle 0 \cdots 0, 1 \rangle \langle 0 \cdots 0, r \rangle.
\]

be the base cycle. The Hamilton cycle of \( R_{j-2} \) is given by

\[
P_j(0,0)\overline{P}_j(0,1)P_j(1,1)\overline{P}_j(1,0).
\]

This implies that we either flip the \( i \)-th bit or add 1 to \( i \) (opposite direction of \( R_r \)) to get the next one when the number of 1's in \( a_1 a_2 \cdots a_{2p} \) is odd. Similarly, if the number of 1's in \( a_1 a_2 \cdots a_{2p} \) is even, we either flip the \( i \)-th bit or subtract 1 for \( i \) (same direction of \( R_r \)). If \( v \) is in the form \( \langle a_1 \cdots a_{i-2} xy \underbrace{0 \cdots 0}_{s \text{ even}}, i \rangle \), \( s \) even, and \( xy = 00 \) or \( xy = 11 \), then we flip the \( i \)-th bit according to the recursive construction. Let \( l = \max\{j : a_j = 1\} \) and \( k = \text{number of 1's in } a_1 \cdots a_{2p} \). Hence, we flip the \( i \)-th bit when \( i \geq l \), and \( i \)
Figure 7.6: Hamilton Cycles in the 4- and 5-dimensional Cube-connected-cycles
and $k$ are both even. If $v$ is in the form $(a_1 \cdots a_{i-1} xy 0 \cdots 0, i)$, $s$ odd, and $xy = 01$ or $xy = 10$, then again we flip the $i$th bit according to the recursive construction. Hence, we flip the $i$th bit when $i \geq l - 1$, and $i$ and $k$ are both odd. We have the following algorithm for the even case.

**Algorithm 7.3**: Algorithm for generating a Hamilton cycle of $CCC_r$ when $r$ is even.

The level indices are reduced modulo $r$ and in the range from 1 to $r$.

1. $(w_0, i_0) \leftarrow (00 \cdots 0, r)$.

2. For $p = 1$ to $r^{2^r}$
   
   let $l = \begin{cases} 0 & \text{if } w_{p-1} = 00 \cdots 0 \\ \max\{j : j \text{th bit of } w_{p-1} \text{ is 1}\} & \text{otherwise,} \end{cases}$

   let $k = \text{number of 1's in } w_{p-1}$.

   If $k$ is even then
   
   if $i_{p-1} \geq l$ then
   
   if $i_{p-1}$ is even then
   
   $w_p \leftarrow \text{flip}(w_{p-1}, i_{p-1}), i_p \leftarrow i_{p-1}$,

   else
   
   $w_p \leftarrow w_{p-1}, i_p \leftarrow i_{p-1} - 1$,

   else
   
   $w_p \leftarrow w_{p-1}, i_p \leftarrow i_{p-1} - 1$,

   else
   
   if $i_{p-1} \geq l - 1$ then
   
   if $i_{p-1}$ is odd then
   
   $w_p \leftarrow \text{flip}(w_{p-1}, i_{p-1}), i_p \leftarrow i_{p-1}$,

   else
   
   $w_p \leftarrow w_{p-1}, i_p \leftarrow i_{p-1} + 1$,

   else
   
   $w_p \leftarrow w_{p-1}, i_p \leftarrow i_{p-1} + 1$.

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3. The Hamilton cycle is
\[(w_0, i_0)(w_1, i_1) \cdots (w_{r2r}, i_{r2r}).\]

The odd case is basically the same as the even case. Since the recursive construction starts from \(R_{r-3}\), some modifications are needed.

**Algorithm 7.3:** Algorithm for generating a Hamilton cycle of \(CCC_r\) when \(r\) is odd.

The level indices are reduced modulo \(r\) and in the range from 1 to \(r\).

1. \((w_0, i_0) \leftarrow (00 \cdots 0, r)\)

2. For \(p = 1\) to \(r2^r\)
   
   \[
   l = \begin{cases} 
   0 & \text{if } w_{p-1} = 00 \cdots 0 \\
   \max\{j : j\text{th bit of } w_{p-1} \text{ is } 1\} & \text{otherwise,}
   \end{cases}
   \]

   let \(k = \) number of 1’s in \(w_{p-1}\).

   If \(k\) is even then
   
   if \(a_{r-1} = 1\) then
   
   if \(i_{p-1} = r - 1\) then
   
   \[
   w_p \leftarrow \text{flip}(w_{p-1}, i_{p-1}), \quad i_p \leftarrow i_{p-1},
   \]
   
   else
   
   \[
   w_p \leftarrow w_{p-1}, \quad i_p \leftarrow i_{p-1} + 1,
   \]
   
   else
   
   if \(l \geq r - 1\) then
   
   if \(i_{p-1} = r - 1\) then
   
   \[
   w_p \leftarrow \text{flip}(w_{p-1}, i_{p-1}), \quad i_p \leftarrow i_{p-1},
   \]
   
   else
   
   \[
   w_p \leftarrow w_{p-1}, \quad i_p \leftarrow i_{p-1} - 1,
   \]
   
   else
   
   if \(i \geq l\) and \(i\) is even then
   
   \[
   w_p \leftarrow \text{flip}(w_{p-1}, i_{p-1}), \quad i_p \leftarrow i_{p-1},
   \]

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else
\[ wp \leftarrow wp_{-1}, \; ip \leftarrow ip_{-1} - 1, \]
else
if \( a_{r-1} = 1 \) then
if \( ip_{-1} = r \) then
\[ wp \leftarrow flip(wp_{-1}, ip_{-1}), \; ip \leftarrow ip_{-1}, \]
else
\[ wp \leftarrow wp_{-1}, \; ip \leftarrow ip_{-1} - 1, \]
else
if \( l \geq r - 2 \) then
if \( ip_{-1} = r - 2 \) then
\[ wp \leftarrow flip(wp_{-1}, ip_{-1}), \; ip \leftarrow ip_{-1}, \]
else
\[ wp \leftarrow wp_{-1}, \; ip \leftarrow ip_{-1} + 1, \]
else
if \( l - 1 \leq i \leq r - 1 \) and \( i \) is odd then
\[ wp \leftarrow flip(wp_{-1}, ip_{-1}), \; ip \leftarrow ip_{-1}, \]
else
\[ wp \leftarrow wp_{-1}, \; ip \leftarrow ip_{-1} + 1. \]

3. The Hamilton cycle is
\[ \langle w_0, i_0 \rangle \langle w_1, i_1 \rangle \cdots \langle w_{r2}, i_{r2} \rangle. \]

7.5 Connectivity

Since the cube-connected-cycles has the bounded degree, the connectivity is bounded. Using the fact that the cube-connected-cycles is hamiltonian, we have the following result.
Theorem 7.8 The connectivity of the r-dimensional cube-connected-cycles is 3.

Proof: Let \( u, v \) be any two vertices in \( CCC_r \). Since \( CCC_r \) is vertex-transitive, we can assume that \( u = (0 \cdots 0, r) \). Using the Algorithm 7.3, we can get a Hamilton cycle \( (0 \cdots 0, r) \triangleq (0 \cdots 0, 1) \). If \( A \) or \( B \) is empty, then \( CCC_r - \{(0 \cdots 0, r), v\} \) is connected. We assume that both \( A \) and \( B \) are non-empty. Then \( (0 \cdots 0, 1) \in A \) and \( (0 \cdots 0, 1) \in B \). Consider the paths

\[ P_1 : (0 \cdots 0, 1)(0 \cdots 0, 1)(0 \cdots 0, 1) \]
\[ (1 \cdots 0, 1)(1 \cdots 0, 0, 1), \]
\[ P_2 : (0 \cdots 0, 1)(0 \cdots 0, 1, r - 1)(0 \cdots 0, 1, 2)(0 \cdots 0, 1, 2) \]
\[ (0 \cdots 0, 1, 3)(0 \cdots 0, 1, r)(0 \cdots 0, 1, r)(0 \cdots 0, 1, r)(0 \cdots 0, r) \]
\[ (0 \cdots 0, r - 1)(0 \cdots 0, 2)(0 \cdots 0, 2)(0 \cdots 0, 1). \]

These are the vertex-disjoint paths from \( (0 \cdots 0, 1) \) to \( (0 \cdots 0, 1) \) without the vertex \( (0 \cdots 0, r) \). Since \( v \) can only disconnect one of the paths, there must exist an edge from \( A \) to \( B \). That is, \( CCC_r - \{(0 \cdots 0, r), v\} \) is connected. Since \( CCC_r \) is 3-regular, the result follows.

Corollary 7.9 The edge-connectivity of the r-dimensional cube-connected-cycles is 3.

Proof: The r-dimensional cube-connected-cycles is 3-regular and has connectivity 3. The result follows.

7.6 Summary

Before finishing the thesis, we make a little summary of the results that we get in Table 7.1.
Table 7.1: Summary of the Results

<table>
<thead>
<tr>
<th>Symbol Set</th>
<th>Transitivity</th>
</tr>
</thead>
<tbody>
<tr>
<td>{(1,2), (3,4), \ldots, (r-1, r)}</td>
<td>distance-transitive</td>
</tr>
<tr>
<td>{(e, 1), (e, r - 1), ((1,2), 1), ((2r-1, 2r), r-1)}</td>
<td>vertex-transitive</td>
</tr>
<tr>
<td>{(e, 1), (e, r - 1), ((1, 2), 0)}</td>
<td>vertex-transitive</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Cayley Graph</th>
<th>(Q_r)</th>
<th>(B_r)</th>
<th>(CCC_r)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Group</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Group</td>
<td>{(1,2), (3,4), \ldots, (r-1, r)}</td>
<td>the group defined in Proposition 6.3</td>
<td>the group defined in Proposition 6.3</td>
</tr>
<tr>
<td>Symbol Set</td>
<td>{(1,2), (3,4), \ldots, (r-1, r)}</td>
<td>{(e, 1), (e, r - 1), ((1,2), 1), ((2r-1, 2r), r-1)}</td>
<td>{(e, 1), (e, r - 1), ((1, 2), 0)}</td>
</tr>
<tr>
<td>Transitivity</td>
<td>distance-transitive</td>
<td>vertex-transitive</td>
<td>vertex-transitive</td>
</tr>
<tr>
<td>Degree</td>
<td>(r)</td>
<td>4</td>
<td>3</td>
</tr>
<tr>
<td>Girth</td>
<td>4</td>
<td>4 if (r \geq 4)</td>
<td>(\min{8, r})</td>
</tr>
<tr>
<td>Girth</td>
<td>3 if (r = 3)</td>
<td></td>
<td>(r)</td>
</tr>
<tr>
<td>Diameter</td>
<td>(r)</td>
<td>(\frac{3r}{2})</td>
<td>(\frac{5r-4}{2})</td>
</tr>
<tr>
<td>Hamiltonian</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Connectivity</td>
<td>(r)</td>
<td>4</td>
<td>3</td>
</tr>
<tr>
<td>Edge-connectivity</td>
<td>(r)</td>
<td>4</td>
<td>3</td>
</tr>
<tr>
<td>Bipartite</td>
<td>all (r)</td>
<td>(r) even</td>
<td>(r) even</td>
</tr>
<tr>
<td>Hamilton-laceable/connected</td>
<td>Yes</td>
<td>Yes</td>
<td>†</td>
</tr>
</tbody>
</table>

† As far as the author knows, the 3-dimensional cube-connected-cycles is hamilton-connected and 4-dimensional cube-connected-cycles is hamilton-laceable.
Bibliography


