



HAL
open science

Graph Problems arising from Wavelength-routing in All-optical Networks

Bruno Beauquier, Jean-Claude Bermond, Luisa Gargano, Pavol Hell,
Stéphane Pérennes, Ugo Vaccaro

► **To cite this version:**

Bruno Beauquier, Jean-Claude Bermond, Luisa Gargano, Pavol Hell, Stéphane Pérennes, et al.. Graph Problems arising from Wavelength-routing in All-optical Networks. WOCS 1997 - 2nd Workshop on Optics and Computer Science - 11th International Parallel Processing Symposium, Apr 1997, Geneve, Switzerland. hal-03764663

HAL Id: hal-03764663

<https://inria.hal.science/hal-03764663>

Submitted on 30 Aug 2022

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

Graph Problems arising from Wavelength–Routing in All–Optical Networks

Bruno Beauquier*
beauquie@essi.fr

Jean-Claude Bermond*
bermond@unice.fr

Luisa Gargano†
lg@dia.unisa.it

Pavol Hell‡
pavol@cs.sfu.ca

Stephane Perennes*
sp@essi.fr

Ugo Vaccaro†
uv@dia.unisa.it

Abstract

This paper surveys the theoretical results obtained for wavelength–routing all–optical networks, presents some new results and proposes several open problems. In all–optical networks the vast bandwidth available is utilized through *wavelength division multiplexing*: a single physical optical link can carry several logical signals, provided that they are transmitted on different wavelengths. The information, once transmitted as light, reaches its destination without being converted to electronic form inbetween, thus reaching high communication speed. We consider both networks with arbitrary topologies and particular networks of practical interest.

1 Introduction

Motivation. Optical networks offer the possibility of interconnecting hundreds to thousands of users, covering local to wide area, and providing capacities exceeding those of traditional technologies by several orders of magnitude. Traditional networks use the electrical form to switch signals which can be modulated electronically at a maximum bit rate of the order of 10 Gbps, while the optical fiber bandwidth is about 10 THz [42]. Thus optical networks are orders of magnitude faster than traditional networks.

Optics is thus emerging as a key technology in state of the art communication networks and is expected to dominate many applications, such as video conferencing, scientific visualisation and real-time medical imaging, high–speed super-computing and distributed computing [20, 38, 47]. We refer to the books of Green [20] and McAulay [28] for a comprehensive overview of the physical theory and applications of this emerging technology.

All–optical communications networks (or *single–hop* networks, see [29]) exploit photonic technologies for the implementation of both switching and transmission functions [19]. In such networks the information, once transmitted as light, reaches its final destination directly without being converted to electronic form inbetween. Maintaining the signal in optical form allows for high speed data transmission in these networks since there is no prohibitive overhead due to

*SLOOP, joint project I3S-CNRS/UNSA/INRIA, 930 Route de Colles, BP 145, 06903 Sophia Antipolis Cedex, France. Work partially supported by the French GDR/PRC Project PRS and by Galileo Project.

†Dipartimento di Informatica ed Applicazioni, Università di Salerno, 84081 Baronissi (SA), Italy. Work partially supported by the MURST in the framework of the 40% Project: Algoritmi, Modelli di Calcolo e Strutture Informative, and by Galileo Project.

‡Simon Fraser University, Burnaby, B.C. V5A1S6, Canada.

conversions to and from the electronic form. Such an approach allows thus the elimination of the “electronic bottleneck” of communications networks with electronic switching.

It is widely accepted that the *wavelength-division multiplexing* (WDM) [10] approach provides means to realize high-capacity networks, by partitioning the optical bandwidth into a large number of channels whose speeds match those of the electronic transmission [6]. It allows multiple data streams to be transferred concurrently along the same optical fiber.

The Optical Model. In general, a WDM optical network consists of routing nodes interconnected by point-to-point fiber-optic links, which can support a certain number of wavelengths. The same wavelength on two input ports *cannot* be routed to a same output port, due to the electromagnetic interference. In this paper we consider *switched* networks with *generalized* switches based on acousto-optic filters [9], as is done in [1, 3, 40]. In this kind of networks, signals for different requests may travel on a same communication link into a node v (on different wavelengths) and then exit v along different links, keeping their original wavelength. Thus the photonic switch can differentiate between several wavelengths coming along a communication line and direct each of them to a different output of the switch. The only constraint on the solution is that no two paths in the network sharing the same optical link have the same wavelength assignment. In switched networks it is possible to “reuse wavelengths” [38], thus obtaining a drastic reduction on the number of required wavelengths with respect to switchless networks [1].

A switched optical network consists of interconnected nodes which can be terminals, switches, or both. Terminals send and receive signals, and switches direct their input signals to one or more of the output links. Each link is bidirectional and actually consists of a pair of unidirectional links [38].

Some authors considered topologies with single undirected fiber links carrying undirected paths ([35, 1, 40, 3]). However, it has since become apparent that optical amplifiers placed on the fiber will be directed devices. Hence we model the optical network as a symmetric directed graph $G = (V(G), A(G))$, where each arc represents a point-to-point unidirectional fiber-optic link. A *request* consists of an ordered pair of nodes, and an *instance* of a set of requests. A *solution* consists of settings for the switches in the network, and an assignment of *wavelengths* to the requests, so that there is a directed path (dipath) between the nodes of each request, and that no arc will carry two different signals on the same wavelength. For our purpose, a wavelength will be an integer in the interval $[1, w]$ for some positive integer w .

The cost and feasibility of switching and amplification devices depend on the number of wavelengths they handle. One should be aware of the severe limitations that current optical technologies impose on the amount of available wavelengths per fiber. While experimental systems report large number of up to 100 wavelengths per fiber [34] current state of the art manufacturing processes restrict the number of wavelengths per fiber of commercial WDM multiplexers to as low as 4 (Pirelli), 8 (Lucent Technologies), and up to 20 (IBM). Thus our aim is to minimize the number w of wavelengths used in a solution. If this number required to realize an all-optical process in one round is greater than the available number of wavelengths, then several all-optical rounds are accomplished.

We remark that optical switches do not modulate the wavelengths of the signals passing through them. If an intermediate node could change the wavelength on which a signal is transmitted, routing an instance using the minimum number of wavelengths would be equivalent to the integer multicommodity flow problem. Unfortunately, current or foreseeable technologies cannot implement such a photonic switch.

While, in WDM technology a fiber link requires different wavelengths for every transmission, SDM (space division multiplexing) technology allows parallel links for a single wavelength, at an additional cost. The two technologies are combined in practice to find an efficient trade off between the two approaches. However, we focus our study on the directed WDM model. In fact, obtaining good bounds on the number of wavelengths provides additional evidence in

favour of the WDM approach [30].

The actual process of setting up switches and routes, and of assigning wavelengths, is done using an electronic backbone control network. We may wonder at the use of a relatively slow electronic network to set up these high-speed connections. In fact, the major applications for such networks require connections that last for relatively long periods once set up. Thus the initial overhead is acceptable as long as sustained throughput at high data rates is subsequently available.

Content of the paper. In this paper we survey the main theoretical results in the area of wavelength–routing in all–optical networks. To keep our overview as complete and current as possible, we have included new results not yet published, some of them obtained by different subsets of authors of the present paper. We have also included several open problems and conjectures, which we hope will stimulate further research in the area.

2 Definitions

There are several natural ways in which all–optical networks can be modeled. In this paper, we choose for the most part to model it as a *symmetric digraph*, that is a directed graph, with vertex set $V(G)$ and arc set $A(G)$, such that if $\alpha = (u, v) \in A(G)$ then $\alpha' = (v, u) \in A(G)$. On the other hand, the definitions given below apply to any (symmetric or non-symmetric) digraphs, and in some cases we will make comments about general digraphs. We always denote by N the number of vertices in G , that is $N = |V(G)|$. We also use the following notation:

- $P(x, y)$ denotes a *dipath* in G from the node x to y , that is a directed path which consists of a set of consecutive arcs beginning in x and ending in y .
- $\delta(x, y)$ denotes the *distance* from x to y in G , that is the length of a shortest dipath $P(x, y)$.
- An algorithm will be said *efficient* if it is deterministic and runs in polynomial time.

Wavelength-routing problem

- A *request* is an ordered pair of nodes (x, y) in G (corresponding to a message to be sent by x to y).
- An *instance* I is a collection of requests. Note that a given request (x, y) can appear more than once in an instance.
- A *routing* R for an instance I in G is a set of dipaths $R = \{P(x, y) \mid (x, y) \in I\}$.
- The *conflict graph* associated to a routing R is the undirected graph (R, E) with vertex set R and such that two dipaths of R are adjacent if and only if they share an arc of G .
- Let G be a digraph and I an instance. The *problem* (G, I) consists of finding a routing R for the instance I and assigning each request $(x, y) \in I$ a wavelength, so that no two dipaths of R sharing an arc have the same wavelength. If we think of wavelengths as colors, the problem (G, I) seeks a routing R and a vertex coloring of the conflict graph (R, E) , such that two adjacent vertices are colored differently. We denote by $\vec{w}(G, I, R)$ the chromatic number of (R, E) , and by $\vec{w}(G, I)$ (or briefly just \vec{w} if there is no ambiguity) the smallest $\vec{w}(G, I, R)$ over all routings R . Thus $\vec{w}(G, I, R)$ is the minimum number of wavelengths for a routing R and $\vec{w}(G, I)$ a minimum number of wavelengths of any routing for (G, I) .

Remark 1 As explained in the introduction, early models of all-optical networks used undirected graphs, and many results are formulated in that context. In this ‘undirected’ model, to conflict means to share an edge. All the definitions for the directed case have natural analogues in the undirected case. Note that we use the same notation G for two different objects: an undirected graph or its induced symmetric digraph obtained by replacing each edge by two opposite arcs. However, in what follows the difference will be made clear by the use of an arrow above the parameters in the directed case.

Remark 2 Any routing by undirected paths induces a routing by directed paths, and a coloring of the undirected paths is also a coloring of the directed paths, as two edge-disjoint paths will become arc-disjoint dipaths. Hence $\vec{w}(G, I) \leq w(G, I)$ for any problem (G, I) and every upper bound on w is an upper bound on \vec{w} . At first glance one could think that $w \leq 2\vec{w}$. That is not true as shown by the case of the instance $I = \{(1, 2), (2, 3), (3, 1)\}$ in the 3-star network G where $V(G) = \{0, 1, 2, 3\}$ and $E(G) = \{\{0, 1\}, \{0, 2\}, \{0, 3\}\}$. Indeed $\vec{w}(G, I) = 1$ and $w(G, I) = 3$. Furthermore it is not hard to note that the ratio w/\vec{w} may be arbitrarily great: by considering the class of mesh-like networks given in [1], we can have for every natural number n , $\vec{w} = 1$ and $w = n$.

Special instances

- The *All-to-All* instance is $I_A = \{(x, y) \mid x \in V(G), y \in V(G), x \neq y\}$.
- A *One-to-All* instance is a set of requests $I_0 = \{(x_0, y) \mid y \in V(G), y \neq x_0\}$, where $x_0 \in V(G)$. A *One-to-Many* instance is a subset of some instance I_0 .
- A *k-relation* is an instance I_k in which each node is a source and a destination of no more than k requests. Hence a permutation instance is a 1-relation and the instance I_A is an $(N - 1)$ -relation.

3 A related parameter

- Given a network G and a routing R for an instance I , the *load* of an arc $\alpha \in A(G)$ in the routing R , denoted by $\vec{\pi}(G, I, R, \alpha)$, is the number of dipaths of R containing α . The *load* of G in the routing R (also called *congestion*), denoted by $\vec{\pi}(G, I, R)$, is the maximum load of any arc of G in the routing R , that is $\vec{\pi}(G, I, R) = \max_{\alpha \in A(G)} \vec{\pi}(G, I, R, \alpha)$.
- The *load* of G for an instance I , denoted by $\vec{\pi}(G, I)$ or $\vec{\pi}$ if there is no ambiguity, is the minimum load of G in any routing R for I , that is $\vec{\pi}(G, I) = \min_R \vec{\pi}(G, I, R)$. For the All-to-All instance I_A , $\vec{\pi}(G, I_A)$ (respectively $\pi(G, I_A)$) is called the *arc forwarding index* (resp. *edge forwarding index*, see [22, 41]) of G .

The relevance of this parameter to our problem is shown by the following lemma:

Lemma 3 $\vec{w}(G, I) \geq \vec{\pi}(G, I)$ for any instance I in any network G .

In other words, to solve a given problem (G, I) one has to use a number of wavelengths at least equal to the maximum number of dipaths having to share an arc. The inequality can be strict, as shown by Figure 1, analogous to an example from [30].

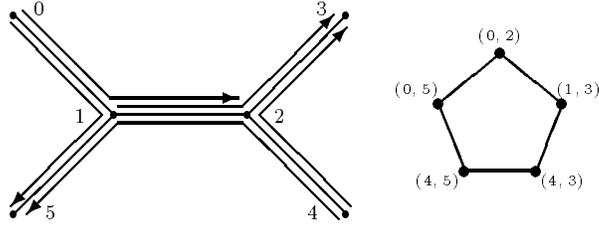


Figure 1: A routing for five requests in a tree and its associated conflict graph.

Indeed for this instance I in the tree G , the load is $\vec{\pi}(G, I) = 2$ but $\vec{w}(G, I) = 3$, since the conflict graph is a pentagon which has chromatic number 3 and clique number 2.

In general, minimizing the number of wavelengths is not the same problem as that of realizing a routing that minimizes the number of dipaths sharing a same arc. Indeed, our problem is made much harder due to the further requirement of wavelengths assignment on the dipaths. In order to get equality in Lemma 3, one should find a routing R such that $\vec{\pi}(G, I, R) = \vec{\pi}(G, I)$, for which the associated conflict graph is $\vec{\pi}(G, I)$ -vertex colorable.

Question 4 Does there always exist a routing R such that $\vec{\pi}(G, I, R) = \vec{\pi}(G, I)$ and at the same time $\vec{w}(G, I, R) = \vec{w}(G, I)$?

Theorem 5 *Determining $\vec{\pi}(G, I)$ for general problems (G, I) is NP-complete.*

Sketch of proof. We first observe that the determination of $\vec{\pi}(G, I)$ is equivalent to the multicommodity integral flow (MIF) directed problem with constant capacities. The theorem holds due to the fact that the (MIF) directed problem with two commodities and capacities all equal to one is shown to be NP-complete in [13]. \square

For some special problems, the determination of $\vec{\pi}(G, I)$ can be efficiently solved. That is obviously the case for trees, where we always have a unique routing. That is also the case of the One-to-Many instances where the problem can be reduced to a flow problem (in the graph obtained from G by considering the sender node as the source, giving a capacity $\vec{\pi}$ to each arc of G , and joining all the vertices of G to a sink t with arcs of capacity 1).

Remark 6 We can define analogously the load $\pi(G, I)$ for an undirected graph and we can prove that $\vec{\pi}(G, I) \leq \pi(G, I) \leq 2\vec{\pi}(G, I)$. For One-to-Many instances I , one can also show that $\vec{\pi}(G, I) = \pi(G, I)$.

Question 7 Does the equality $\vec{\pi}(G, I_A) = \lceil \pi(G, I_A)/2 \rceil$ always hold ?

The following question was also asked in [22]:

Question 8 What is the complexity of determining $\vec{\pi}(G, I_A)$?

Let the *arc expansion* $\beta(G)$ of a directed graph G having N nodes be the minimum, over all subsets of nodes $S \subset V(G)$ of size $|S| \leq N/2$, of the ratio of the number of arcs with origin in S and destination outside of S , to the size of S . It follows from [41] that $\vec{\pi}(G, I_A) \geq \frac{N}{2\beta(G)}$.

4 Arbitrary Networks

4.1 Arbitrary instances

For a general network G and an arbitrary instance I , the problem of determining $\vec{w}(G, I)$ has been proved to be NP-hard in [11]. In particular, in [11] it has been proved that determining $\vec{w}(G, I)$ is NP-hard for trees and cycles. In [12] these results have been extended to binary trees and meshes. NP-completeness results in the undirected model were known much earlier (actually, well before the advent of the WDM technology). In particular, in [17] it is proved that the problem of determining $w(G, I)$ is NP-complete for trees. This result has been extended in [11] to cycles, while in [12] it has been proved that the problem is efficiently solvable for bounded degree trees.

In view of this last result and of the NP-hardness of determining $\vec{w}(G, I)$ for binary trees, it might seem that the problem of computing $\vec{w}(G, I)$ is harder than to compute $w(G, I)$. This is not true in general. For instance, the determination of $w(G, I)$ remains NP-complete when G is a star network, whereas $\vec{w}(G, I)$ can be efficiently computed. Indeed, in the undirected model this problem corresponds to edge-coloring a multigraph [23], each node of which corresponds to a branch in the star network. In the directed case, the same problem becomes equivalent to edge-coloring a bipartite multigraph, efficiently solvable by Hall's theorem.

Let R be a routing for an instance I in a network G . Let L be the maximum length of its dipaths and Δ the maximum degree of its conflict graph. It is clear that $\Delta \leq L\vec{\pi}(G, I, R)$. By a greedy coloring we know that $(\Delta + 1)$ wavelengths are sufficient to solve the problem (G, I) . Thus $\vec{w} = O(L\vec{\pi})$ and similarly $w = O(L\pi)$. A set of critical undirected problems reaching asymptotically this upper bounds has been given in mesh-like networks (see [1]). By adapting their examples (orientating alternatively the vertical links up and down), we can obtain the same result in (non symmetric) digraphs:

Theorem 9 *For every π and L , there exists a problem (G, I) such that $\vec{\pi}(G, I) = \pi$, $L = \max_{(x,y) \in I} \delta(x, y)$ and $\vec{w}(G, I) = \Omega(\pi L)$.*

Question 10 Does Theorem 9 hold for symmetric digraphs ?

4.2 Permutations and k -relations

By deriving a lower bound on the number of links used in the worst case and an upper bound on the total number of links in the network, Pankaj [35, 36] obtained in his thesis some results in the undirected model that are easy to translate in the directed model:

Theorem 11 *For every symmetric digraph G of maximum degree Δ , there exists a worst case permutation instance I_1 such that*

$$\vec{w}(G, I_1) \geq \frac{\lfloor \log_{\Delta} \frac{N}{2} \rfloor}{2\Delta} .$$

Theorem 12 *For every vertex transitive symmetric digraph G of diameter D and degree Δ , there exists a worst case permutation instance I_1 such that*

$$\vec{w}(G, I_1) \geq \left\lceil \frac{D}{\Delta} \right\rceil .$$

In addition, Pankaj obtained the lower bound $(\min\{k, N/2\} \lceil \log_{\Delta} \frac{N}{2} \rceil) / 2\Delta$ for a worst case k -relation I_k . In terms of growth rate, these results show that the necessary number of wavelengths is $\Omega(\frac{\log N}{\Delta \log \Delta})$ for a worst case permutation I_1 and $\Omega(\frac{N \log N}{\Delta \log \Delta})$ for the All-to-All instance I_A .

In the undirected model, Raghavan and Upfal have shown in [40] an existential lower bound which relates the number of wavelengths and the edge expansion, by starting from the same example that we can find in [1]. In the same way that we have obtained Theorem 9, we obtain an existential lower bound for digraphs:

Theorem 13 *For every $\beta \leq 1$ and $1 \leq k \leq N$, there exists a problem (G, I_k) for a k -relation I_k in a digraph G with arc expansion β , such that*

$$\vec{w}(G, I_k) = \Omega(k/\beta^2).$$

Question 14 Does Theorem 13 hold for symmetric digraphs ?

In order to find a routing for an instance I in a network G , one has to choose a dipath for each pair of nodes in I and use these dipaths concurrently. This corresponds to an integer multicommodity flow problem. By using the same arguments as in [3] for permutations, we obtain the following result in the directed model:

Theorem 15 *There is an efficient algorithm to solve the problem (G, I) for every k -relation I_k in any bounded degree network G , using at most $O(k \log^2 N / \beta^2)$ wavelengths.*

Note that this result almost matches the $\Omega(k/\beta^2)$ existential lower bound of Theorem 13. We can also put $\vec{w}(G, I)$ in relation with the *arc connectivity* λ of G . A digraph G has arc connectivity λ if the minimum number of arcs to remove in order to disconnect G is λ . Using a theorem by Shiloach [43], we can prove the following result:

Theorem 16 *There is an efficient algorithm to solve the problem (G, I) for every instance I in any symmetric digraph G with arc connectivity λ , using at most $\lceil \frac{|I|}{\lambda} \rceil$ wavelengths. Moreover, this bound is best possible for worst case instances.*

4.3 Other specific instances

The following theorem gives the exact value of $\vec{w}(G, I_0)$ for a worst case instance I_0 in various classes of important networks, namely the *maximally arc connected* digraphs, including the wide class of vertex transitive digraphs. A digraph G is maximally arc connected if its minimum degree is equal to its arc connectivity.

Theorem 17 (Bermond et al. [8]). *For a worst case One-to-All instance I_0 in a maximally arc connected digraph G of minimum degree $d(G)$,*

$$\vec{w}(G, I_0) = \vec{\pi}(G, I_0) = \left\lceil \frac{N-1}{d(G)} \right\rceil.$$

In addition, an efficient network flow based algorithm is given to solve the problem (G, I) with $\vec{w}(G, I)$ wavelengths, for any One-to-Many instance I in any network G . We have recently generalized a part of the last theorem:

Theorem 18 $\vec{w}(G, I) = \vec{\pi}(G, I)$, for any One-to-Many instance I in any digraph G .

We have been wondering about the following question:

Question 19 Does the equality $\vec{w}(G, I_A) = \vec{\pi}(G, I_A)$ hold for the All-to-All instance I_A in every symmetric digraph G ?

5 Specific networks

5.1 Trees

The case of trees is particularly interesting as many practical networks, e.g., in the telecommunications industry, have a tree-like structure (see [30]). Let us consider first the case when the network G is a symmetric subdivided star, that is, when G is a symmetric tree with at most one node with outdegree greater than 2. In this case, $\vec{w}(G, I) = \vec{\pi}(G, I)$ for every instance I . Actually, when G is a path this is equivalent to the fact that the chromatic number of an interval graph is equal to the maximum size of its cliques. When G is a star this is equivalent to the fact that in a bipartite graph the edge chromatic index is equal to its maximum degree. When G is a subdivided star, we can combine these two approaches. It is not difficult to observe that the converse also holds, i.e., when T is a symmetric tree other than a subdivided star, then there exists an instance I such that $\vec{w}(G, I) \neq \vec{\pi}(G, I)$:

Theorem 20 *Let G be a symmetric tree. Then $\vec{w}(G, I) = \vec{\pi}(G, I)$ for all instances I if and only if G is a subdivided star.*

In fact, a tree other than a subdivided star is a subdivision of the graph shown in Figure 1, thus we can always choose requests such that the conflict graph is a pentagon.

Conjecture 21 (Mihail, Kaklamanis and Rao [30]). *For every instance I in any tree G ,*

$$\vec{w}(G, I) \leq \frac{3\vec{\pi}(G, I)}{2}.$$

Theorem 22 (Kaklamanis and Persiano [25]). *There is an efficient algorithm to solve the problem (G, I) for every instance I in any tree G , using at most $5\vec{\pi}(G, I)/3$ wavelengths.*

We now show that in a tree network there exist problems with arbitrarily great load $\vec{\pi}$, such that the ratio $\vec{w}/\vec{\pi}$ is greater than $5/4$. We start from the problem (G, I) shown in Figure 1. For every natural number n , let I^n denote the instance made of n copies of I (each request of I is repeated n times). Coloring the conflict graph associated to the problem (G, I^n) becomes then a multicoloring problem for the pentagon, and from what is known about this problem (see [21]), we obtain:

Theorem 23 *For every π there exists a problem (G, I) in a tree G , such that $\vec{\pi}(G, I) \geq \pi$ and $\vec{w}(G, I) \geq 5\vec{\pi}(G, I)/4$.*

Question 24 Can the constant $5/4$ of Theorem 23 be raised ?

Note that Theorem 22 implies an approximation algorithm to solve the problem (G, I) in a symmetric tree G with at most $5\vec{w}(G, I)/3$ wavelengths. For undirected graphs, there is a better result:

Theorem 25 (Erlebach and Jansen [11]). *There is an efficient algorithm to solve the problem (G, I) for any instance I in any tree network G , using at most $\lceil 1.1\vec{w}(G, I) + 0.8 \rceil$ wavelengths.*

In the case of subdivided stars above the problem of determining \vec{w} can be efficiently solved, since that is the case for the chromatic number of interval graphs and the chromatic index of bipartite graphs.

In the case of undirected trees G , Tarjan [44] and independently Raghavan and Upfal [40] proved that $w(G, I) \leq 3\pi(G, I)/2$, and this bound is achieved in the example of Remark 2. Edge-coloring of multigraphs is an NP-complete problem [23], and since each multigraph is the conflict graph of some instance I on some star G , the computation of w in stars (and hence trees) is NP-complete.

Theorem 26 (Gargano, Hell and Perennes [16]). *There is an efficient algorithm to solve the problem (G, I_A) for the All-to-All instance I_A in any tree G , using $\vec{w}(G, I_A) = \vec{\pi}(G, I_A)$ wavelengths.*

In the undirected model, there are examples where the ratio $w(G, I_A)/\pi(G, I_A)$ can tend asymptotically to $3/2$. For instance, it is the case for the family of trees having three branches of equal size.

Question 27 Can the problem (G, I_A) for any undirected tree G be efficiently solved, using exactly $w(G, I_A)$ wavelengths ?

To our knowledge, the particular case of permutation instances in tree networks has not been studied in the literature. By adapting the example shown in Figure 1, it follows that we can have $\vec{w}(G, I) = 3\vec{\pi}(G, I_1)/2$ for a permutation instance I_1 in a tree network G .

5.2 Rings, tori and meshes

Theorem 28 (Frank et al. [14]). *There is a linear time algorithm to find for every instance I in any undirected ring network G a routing R for I , such that $\pi(G, I, R) = \pi(G, I)$.*

Question 29 Does Theorem 28 also hold in the directed model ?

A routing being fixed, the wavelength assignment becomes in both models a vertex coloring of circular arc graph, which in the general case is proved to be NP-complete in [15]. As observed earlier, also the general problem in both models of determining \vec{w} or w is NP-hard for ring networks (see [11]). Nevertheless there are some approximation results.

Given a routing R for an instance I in a ring network G , Tucker [46] gave an efficient algorithm to solve the wavelengths assignment problem, using at most $(2\vec{\pi}(G, I, R) - 1)$ wavelengths. Combined with Theorem 28, this result gives an efficient approximation algorithm of ratio two for the problem (G, I) in undirected ring networks G . Using the same idea as Tucker, such approximation algorithms have been shown in the undirected model in [40] and in the directed model in [30]. In addition, Tucker showed examples with arbitrarily great $\vec{\pi}$ necessitating the use of $(2\vec{\pi} - 1)$ wavelengths. For instance, it is the case for the five distinct requests $(x, x + 2 \bmod 5)$ in the 5-ring network.

The following result gives the first *per-instance* approximation algorithm for bounded dimension meshes and it also holds in the directed model and for bounded dimension tori.

Theorem 30 (Aumann and Rabani [3]). *There is an efficient algorithm to solve the problem (G, I) for any instance I in any bounded dimension mesh network G , using at most $O(\log N \log |I| w(G, I))$ wavelengths.*

Rabani [37] improved recently the previous approximation results obtained for the square meshes, although the hidden constants are huge:

Theorem 31 (Rabani [37]). *There is an efficient algorithm to solve the problem (G, I) for every instance I in any square mesh network G , using at most $\text{poly}(\log \log N) \cdot \vec{w}(G, I)$ wavelengths, where poly denotes a polynomial function.*

This result also holds in the directed model and it has been given in addition an efficient algorithm to determine \vec{w} in square meshes with a constant approximation ratio.

Theorem 32 (Bermond et al. [8]). *For the All-to-All instance I_A in the ring network G with N nodes, $\vec{w}(G, I_A) = \vec{\pi}(G, I_A) = \left\lceil \frac{1}{2} \left\lfloor \frac{N^2}{4} \right\rfloor \right\rceil$.*

The following theorem extends results from [2]:

Theorem 33 (Beauquier [5]). *For the All-to-All instance I_A in the d -dimensional hypersquare torus G with even side, $\vec{w}(G, I_A) = \vec{\pi}(G, I_A) = N^{\frac{d+1}{d}}/8$.*

5.3 Hypercubes

By using a non-standard embedding of the Beneš network and a variation on the classical routing algorithm, Aumann and Rabani [3] obtained a realization of any permutation in the hypercube with a constant number of wavelengths. This result obtained in the undirected model also holds in the directed model. In comparison, the standard embedding of the Beneš network in the hypercube may require to use $\Omega(\log N)$ wavelengths in the worst case.

Theorem 34 (Aumann and Rabani [3]). *For any permutation instance I_1 in any multidimensional hypercube network G , the problem (G, I_1) can be efficiently solved with 16 wavelengths.*

For the All-to-All instance I_A , the following theorem, independently found by Togni [45], generalizes a result obtained for hypercubes in [35], and independently in [8]:

Theorem 35 (Beauquier [5]). *For the All-to-All instance I_A in any cartesian sum G of complete graphs, the problem (G, I_A) can be efficiently solved with $\vec{w}(G, I_A) = \vec{\pi}(G, I_A)$ wavelengths.*

6 Final Remarks

We have surveyed the main theoretical results arising from wavelength-routing in all-optical networks and posed several questions for future research. We want now just to point out some recent work related in some way to the topic of this paper. For instance, in [32, 33, 26] the following problem is considered: given a graph G , a number of colors and a collection of paths \mathcal{P} in G , color a maximum number of paths in \mathcal{P} such that no two of them sharing an edge have the same color. Of course, this problem coincides with ours for trees. When G is an oriented tree (each edge oriented in exactly one way), Monma and Wei [31] have proved that for any instance I , we have $\vec{w}(G, I) = \vec{\pi}(G, I)$ and $\vec{w}(G, I)$ can be efficiently computed.

Another very important line of research is that of on-line routing in optical networks. In this scenario, requests can dynamically change and are given at different times. We refer to [4] and references therein quoted for an account of this area.

Finally, it is worth pointing out that the results presented are related to the problem of designing *logical topologies* over a wavelength-routing optical network physical topology (see [39] and references therein quoted for an account of this area of research).

References

- [1] A. Aggarwal, A. Bar-Noy, D. Coppersmith, R. Ramaswami, B. Schieber, M. Sudan. “Efficient Routing and Scheduling Algorithms for Optical Networks”, in *Proceedings of the 5th Annual ACM-SIAM Symposium on Discrete Algorithms (SODA '94)*, (1994), 412–423.
- [2] M. Ajmone Marsan, A. Bianco, E. Leonardi and F. Neri. “Topologies for Wavelength–Routing All-Optical Networks”, *IEEE/ACM Transactions on Networking*, vol. 1, No. 5, (1993), 534–546.
- [3] Y. Aumann and Y. Rabani. “Improved Bounds for All Optical Routing”, *Proceedings of the 6th Annual ACM-SIAM Symposium on Discrete Algorithms (SODA '95)*, (1995), 567–576.
- [4] B. Awerbuch, Y. Azar, A. Fiat, S. Leonardi, and A. Rosen, “On–Line Competitive Algorithms for Call Admission in Optical Networks”, *Proceedings of ESA '96*, LNCS 1136, (1996), 431–444.
- [5] B. Beauquier. “All–to–All Communication for some Wavelength–Routing All–Optical Networks”, *Unpublished Manuscript*, (1996).
- [6] C. Brackett. “Dense Wavelength Division Multiplexing Networks’: Principle and Applications, *IEEE J. Selected Areas in Communications*, vol. 8, (1990), 373–380.
- [7] R. A. Barry and P. A. Humblet, “On the Number of Wavelengths and Switches in All–Optical Networks”, in *IEEE Transactions on Communications*, (1993).
- [8] J-C. Bermond, L. Gargano, S. Perennes, A. Rescigno and U. Vaccaro. “Efficient Collective Communications in Optical Networks”, *ICALP '96*, F. Meyer auf der Heide and B. Monien (Eds.), Lectures Notes in Computer Science, vol. 1099, pp. 574–585, Springer-Verlag, 1996.
- [9] K-W. Cheng. “Accousto-Optic Tunable Filters in Narrowband WDM Networks”, *IEEE JSAC*, vol. 8 (1990), 1015–1025.
- [10] N.K. Cheung, K. Nosu and G. Winzer. *IEEE JSAC: Special Issue on Dense WDM Networks*, vol. 8, (1990).
- [11] T. Erlebach and K. Jansen. “Scheduling of Virtual Connections in Fast Networks”, *Proc. of 4th Workshop on Parallel Systems and Algorithms PASA '96*, (1996), 13–32.
- [12] T. Erlebach and K. Jansen. “Call Scheduling in Trees, Rings and Meshes”, to appear in *Hawaii Int. Conf. on Syst. Sciences (HICSS)*, (1997).
- [13] S. Even, A. Itai and A. Shamir. “On the Complexity of Timetable and Multicommodity Flow Problems”, *SIAM Journ. of Computing*, vol. 5, No. 4, (1976), 697–703.
- [14] A. Frank, T. Nishizeki, N. Saito, H. Suzuki and E. Tardos. “Algorithms for routing around a rectangle”, *Discrete Applied Mathematics*, vol. 40, (1992), 363–378.
- [15] M. R. Garey, D. S. Johnson, G. L. Miller and C. H. Papadimitriou. “The Complexity of Coloring Circular Arcs and Chords”, *SIAM Journ. Alg. Disc. Meth.*, vol. 1, No. 2, (1980), 216–227.
- [16] L. Gargano, P. Hell and S. Perennes. “Coloring All Directed Paths in a Symmetric Tree”, *Unpublished Manuscript*, (1996).
- [17] M. C. Golumbic and R. E. Jamison. “The Edge Intersection Graphs of Paths in a Tree”, *Journ. Comb. Theo.*, Series B, vol.38, (1985), 8–22.
- [18] M. C. Golumbic and R. E. Jamison. “Edge and Vertex Intersection of Paths in a Tree”, *Discrete Mathematics*, vol. 55, (1985), 151–159.
- [19] P. E. Green. “The Future of Fiber–Optic Computer Networks”, *IEEE Computer*, vol. 24, (1991), 78–87.
- [20] P. E. Green. *Fiber–Optic Communication Networks*, Prentice–Hall, 1992.
- [21] P. Hell and F.S. Roberts. “Analogues of Shannon capacity”, *Annals of Discrete Math.*, vol. 12, (1982), 155–168.

- [22] M. C. Heydemann, J-C. Meyer and D. Sotteau, "On Forwarding Indices of Networks", *Discrete Applied Mathematics*, vol. 23, (1989), 103–123.
- [23] I. Holyer. "The NP-completeness of Edge Coloring", *SIAM J. of Comp.*, vol. 10, No. 4, (1981), 718–720.
- [24] C. Kaklamanis and P. Persiano, "Efficient Wavelength Routing on Directed Fiber Trees", *Proc. ESA '96*, Springer Verlag, LNCS 1136, (1996), 460–470.
- [25] C. Kaklamanis and P. Persiano. "Constrained Bipartite Edge Coloring with Applications to Wavelength Routing", *Unpublished Manuscript*.
- [26] J. Kleinberg and E. Tardos. "Disjoint Paths in Densely Embedded Graphs", *Proc. of 36th Annual IEEE Symposium on Foundations of Computer Science (FOCS'95)*, (1995), 52–61.
- [27] F. T. Leighton and S. Rao, "An Approximate Max-Flow Min-Cut Theorem for Uniform Multicommodity Flow Problems with Applications to Approximation Algorithms", in: *Proceedings of the 29th Annual IEEE Symposium on Foundations of Computer Science (FOCS'88)*, (1988), 422–431.
- [28] A. D. McAulay. *Optical Computer Architectures*, John Wiley, 1991.
- [29] B. Mukherjee, "WDM-Based Local Lightwave Networks, Part I: Single-Hop Systems", *IEEE Networks*, vol. 6, (1992), 12–27.
- [30] M. Mihail, K. Kaklamanis, S. Rao, "Efficient Access to Optical Bandwidth", *Proc. of 36th Annual IEEE Symposium on Foundations of Computer Science (FOCS'95)*, (1995), 548–557.
- [31] C. L. Monma and V. K. Wei. "Intersection Graphs of Paths in a Tree", *Journal of Combinatorial Theory, Series B*, (1986), 141–181.
- [32] C. Nomikos, "Approximability of the Path Coloring Problem", *Unpublished Manuscript*.
- [33] C. Nomikos and C. Zachos, "Satisfying a Maximum Number of Communication Requests", *Unpublished Manuscript*.
- [34] K. Nosu, H. Toba, K. Inoue, and K. Oda, "100 Channel Optical FDM Technology and its Applications to Optical FDM Channel-Based Networks", *IEEE/OSA JLT*, vol. 11, (1993), 764–776.
- [35] R. K. Pankaj, *Architectures for Linear Lightwave Networks*, PhD Thesis, Dept. of Electrical Engineering and Computer Science, MIT, Cambridge, MA, 1992.
- [36] R. K. Pankaj and R. G. Gallager. "Wavelength Requirements of All-Optical Networks", *IEEE/ACM Transactions on Networking*, vol. 3, (1995), 269–280.
- [37] Y. Rabani. "Path Coloring on the Meshes", *Proc. of FOCS '96*.
- [38] R. Ramaswami, "Multi-Wavelength Lightwave Networks for Computer Communication", *IEEE Communication Magazine*, vol. 31, (1993), 78–88.
- [39] R. Ramaswami and K. Sivaraman, "Design of Logical Topologies for Wavelength-Routed Optical Networks", *IEEE JSAC/JLT Special Issue on Optical Networks*, June 1996.
- [40] P. Raghavan and E. Upfal. "Efficient Routing in All-Optical Networks", *Proceedings of the 26th Annual ACM Symposium on Theory of Computing (STOC'94)*, (1994), 134–143.
- [41] P. Solé, "Expanding and Forwarding", *Discrete Applied Mathematics*, vol. 58, (1995), 67–78.
- [42] M. Settembre and F. Matera. "All Optical Implementations of High Capacity TDMA Networks", *Fiber and Integrated Optics*, vol. 12, (1993), 173–186.
- [43] Y. Shiloach. "Edge-Disjoint Branching in Directed Multigraphs", *Inf. Proc. Letters*, vol. 8, 24–27.
- [44] R. E. Tarjan. "Decomposition by Clique Separators", *Discrete Mathematics*, vol. 55, (1985), 221–232.
- [45] O. Togni. "Echange Total dans les Réseaux Optiques obtenus par Composition de Cliques", *Unpublished Manuscript*, (1996).
- [46] A. Tucker. "Coloring a Family of Circular Arcs", *SIAM J. Appl. Math.*, vol. 29, No. 3, (1975), 493–502.
- [47] R. J. Vetter and D. H. C. Du. "Distributed Computing with High-Speed Optical Networks", *IEEE Computer*, vol. 26, (1993), 8–18.