# Intelligent query processing in P2P networks: semantic issues and routing algorithms

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Abstract. P2P networks have become a commonly used way of disseminating content on the Internet. In this context, constructing efficient and distributed P2P rout-8 ing algorithms for complex environments that include a huge number of distributed 9 nodes with different computing and network capabilities is a major challenge. In the 10 last years, query routing algorithms have evolved by taking into account different 11 features (provenance, nodes' history, topic similarity, etc.). Such features are usu-12 ally stored in auxiliary data structures (tables, matrices, etc.), which provide an extra 13 knowledge engineering layer on top of the network, resulting in an added semantic 14 value for specifying algorithms for efficient query routing. This article examines the 15 main existing algorithms for query routing in unstructured P2P networks in which 16 semantic aspects play a major role. A general comparative analysis is included, as-17 sociated with a taxonomy of P2P networks based on their degree of decentralization 18 and the different approaches adopted to exploit the available semantic aspects. 19

#### **1.** Introduction

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A peer-to-peer network (or just P2P network) is a computing model present in almost 21 every device, from smartphones to large-scale servers, as a way to leverage large amounts 22 of computing power, storage, and connectivity around the world. In a P2P network, each 23 peer can act indistinctly as a client and a server and can collaborate in order to share infor-24 mation in a distributed environment without any centralized coordination. These systems 25 are vulnerable to security problems, abuse, and other threats, and consequently, it is nec-26 essary to be resilient to different forms of attacks, to have mechanisms to detect and re-27 move poisoned data [20,72], and to distinguish spammers from honest peers [92,122,94]. 28 Despite these issues, P2P networks are widely used for large-scale data sharing, content 29 distribution, and application-level multicast working with a tolerable waiting time for the 30 users [86,97,144]. 31

P2P technologies have demonstrated great potential to support distributed informa-32 tion retrieval. The typical information retrieval problem in P2P networks involves finding 33 a set of documents in the network that are relevant to a given query. To better describe the problem of information retrieval in P2P networks, it is useful to identify a number 35 of salient features, as illustrated in Figure 1. In a P2P information retrieval network each 36 device or node (peer) maintains a collection of documents available to share with other 37 peers. In order for those peers to interact with each other, several components are required 38 to support search among peers associated with a given query. These components include 39 a routing algorithm, routing tables, indices, and an established protocol that manages the 40

queries that each node can handle [56,48,121]. The routing algorithm determines how a
 node searches for information by sending query messages to other peers. When a peer re ceives a query message, it attempts to retrieve relevant documents from its own collection,
 forwarding as well the query to other peers in the network.

In the context of P2P networks, it is necessary to distinguish between the *physical* network and the *logical* network. The former consists of real physical connections between devices while the latter is a topology that emerges from the peers' interaction. Since the

8 interaction between peers can be guided by their semantic relations, semantic communi-

<sup>9</sup> *ties* commonly emerge in logical networks [36,22].



Fig. 1. Conceptual elements which characterize a P2P network.

Different methodologies and techniques for information retrieval on P2P networks have been proposed [129], providing alternative approaches to exploit the concept of social communities on the Internet. Some of the benefits which result from applying P2P information retrieval networks are the following:

- By their very nature, P2P networks do not need a centralized administration, being

- self-organizing and adaptive (in the sense that peers can enter and leave the network
   without any external control).
- Peers can have access to several storage and processing resources available from dif ferent computers and devices in the network.
- Since pure P2P networks are distributed and decentralized, they tend to be fault tolerant and with a good load balance for handling network traffic.

Over the years, the Internet has become increasingly restricted to client-server applications. Unfriendly protocols and firewalls are examples of aspects that restrict and limit the use of Internet. To some extent, P2P technologies can be thought of as a way of returning the Internet to its original cooperative design, in which every participant creates as well as consumes [101]. The emerging P2P networks could thus empower almost any group of people with shared interests such as culture, politics, health, etc.

In this article, we present a review of literature solutions to the information retrieval 1 problem in unstructured P2P networks and a description of the main techniques for rout-2 ing queries in structured and semi-structured P2P systems. Our analysis includes a novel 3 classification of these systems, putting special emphasis on their semantic aspects and the different existing routing algorithms. While existing algorithms have facilitated the 5 implementation of robust distributed architectures, there are still several limitations faced by current search mechanisms. Indeed, the information retrieval problem is more com-7 plex than the traditional problem of searching for resources based on object identifiers or names. Over the years, the information retrieval community has developed numerous doc-9 ument retrieval techniques for centralized search. However, these methods cannot be di-10 rectly applied to P2P information retrieval networks, where search is not centralized since 11 documents are distributed among a large number of repositories. Given the information 12 explosion that we have experienced in the last years, such new capabilities are an impor-13 tant step for making P2P networks effective in many applications that go beyond simple 14 data storing. There has been previous research work which provided the background for 15 our analysis: [147] presents a review of some early methods developed to address infor-16 mation retrieval problems in P2P networks. An empirical comparison of some of these 17 methods is presented in [132] and a more recent survey of the major challenges for P2P 18 information retrieval is presented in [129]. 19

The remainder of the article is organized as follows: Section 2 presents some background concepts used in the rest of the paper, including a description of the main components of P2P networks and a classification of P2P search algorithms. Section 3 presents a summary of the most important search algorithms in P2P networks, highlighting those that make use of semantic aspects. Section 4 provides a comparison and classification of the algorithms presented in Section 3. Finally, the conclusions derived from this analysis are presented in Section 5.

## 27 2. Background

According to [117], a P2P network is, in its pure form, "a distributed system in which 28 every peer communicates with other peers without the intervention of centralized hosts". 29 In real-world P2P networks, the participating peers are typically computers to be found 30 at the edge of the network, in people's offices or homes [77], Thus, a P2P network turns 31 out to be formed by a set of machines, which offer a wide range of capabilities when con-32 sidering storage and Internet access speeds, being attractive for different computing tasks 33 (such as file sharing, media streaming, and distributed search). P2P networks have usu-34 ally no centralized directory, being *self-organizing*, with the ability to adapt to different 35 circumstances associated with the participating peers (e.g. joining in, failing or departing 36 from the network). It is worth noticing that the use of a common language ensures that the 37 communication between peers is symmetric for both the provision of services and com-38 munication capabilities. From this symmetry the P2P network can also be characterized 39 as *self-scaling*, since each peer that joins the network adds a new computational resource 40 to the available total capacity [32,112,29]. There are many important challenges specific 41 to P2P networks [45], such as how to administrate resources properly, how to provide an 42 acceptable quality of service while guaranteeing robustness and availability of data, etc. 43 Reviews of different P2P frameworks and their applications can be found in [24,105,89]. 44

The rest of this section will present different dimensions relevant for assessing P2P networks. First, in subsection 2.1 we present a common classification for P2P networks. Then, subsection 2.2 introduces the concept of *semantic aspect* in the context of P2P information retrieval. Subsection 2.3 introduce a novel taxonomy for routing algorithms based on semantic aspects. Subsection 2.4 present the importance of distributed hash tables for the implementation of routing algorithms mainly for structured topologies. To conclude the section, subsection 2.5 introduces some of the salient applications of the P2P technologies.

## 9 2.1. Network classification

A common approach to classify P2P networks is based on their *degree of centralization*,
 which results in three possible alternatives:

Centralized: These P2P networks have a monolithic architecture with a single server that allows transactions between nodes and keeps track of where content is stored.
 For example, *Napster* [100] had a constantly-updated directory hosted in a central location (the Napster website). This system was extremely successful before its legal issues [50]. Clearly, this centralized approach scales poorly and has a single point of failure.

- Decentralized: These are systems where there is no centralized directory, since each 18 peer acts as a client and a server at the same time. Gnutella [111] is an example of 19 this architecture, where the network is formed by nodes that join and leave the sys-20 tem. Several factors motivate the adoption of decentralized networks such as privacy 21 control, availability, scalability, security, and reliability [107]. On the downside, to 22 find a file in a decentralized network a node must query its neighbors. In its basic 23 form, this method is called *Flooding* and is extremely unscalable, generating large 24 loads on the network participants. 25

Hybrid: These systems have no central directory server and therefore can be seen as decentralized networks. However, they have some special peers or super-peers with extra capabilities. *FreeNet* [41,4] is an example of such system and there is a growing interest on this kind of P2P architectures, which supports a hash-table-like interface [110,124,114,149,63].

The network topology is another common classification criterion for P2P networks, allowing to identify two distinctive groups:

Structured: In structured P2P networks the nodes are organized into a specific topol ogy. This organization ensures that any node can search the network for a resource,
 providing as well a good response time. The most common type of structured P2P
 network is based on a distributed hash table (DHT) that provides a lookup service
 similar to a hash table: (key, value) pairs are stored in a DHT, and any participating
 node can efficiently retrieve the value associated with a given key. This approach is
 adopted by *Chord* [124], *Pastry* [114] and *Tapestry* [149], among others.

Unstructured: In unstructured P2P networks no structure is pre-established over the
 network, but rather these networks are formed by nodes that randomly build connections to each other. Because of their lack of structure, unstructured networks are easy

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to build and highly robust. However, a major limitation of these networks is their poor
 effectiveness in finding resources. The simplest algorithm used on unstructured P2P
 networks is based on propagating the query message through all the network, leading
 to a high amount of traffic. Additionally, it is not possible to ensure that search queries
 will be eventually resolved. Some examples of this type of systems are *Gnutella* [111]
 and *KaZaa* [1].

Figure 2 shows a diagram with the classification of P2P systems, identifying the relationships between the different degrees of centralization and the possible topology of the network. Structured topologies are frequently related to centralized or hybrid systems in order to take full advantage of both features. However, unstructured topologies have random connections and in general any peer is equivalent to the others, so that they are strongly associated with the concept of decentralization. The items marked with a star correspond to the central topics on which we will focus on the rest of this survey.



**Fig. 2.** Classification of P2P networks in terms of their degree of centralization and associated topology.

Search methods depend on the underlying network structure. As discussed before, in the case of pure unstructured networks there is no specific pattern for the organization of its nodes, resulting in a random topology. As a consequence, these networks have relatively low search efficiency when contrasted with structured and hybrid ones, as a single query can generate massive amounts of traffic even if the network has a moderate

size [112]. Thus, many alternatives have been studied to improve the basic flooding approach in unstructured networks, such as random walk [88,71,145,28,80], directing the
search towards potentially useful nodes [21,147,138,120], or clustering peers by content [44,133,33] or interest [123,93,108].

## **5 2.2. Semantic Aspects**

A possible solution to improve communication overhead and scalability in large-scale
unstructured P2P systems is to forward queries to a group of peers that are known to
be potentially useful to answer the query. The selection of potentially useful peers is
typically based on the peers' past activity or their semantic similarity to the original
query [43,123,135,74,85,78,130]. Identifying peers based on their potential to answer a
query in a useful way requires associating semantic aspects with peers.

A *semantic aspect*, in the context of a P2P network, is a feature or a set of features that allows recognizing the semantic of the data stored in a node. Semantic aspects can be exploited by routing algorithms to help peers predict which other peers have knowledge useful to respond to a query. Topical information, past experience, and node-state information are examples of such semantic aspects.

In algorithms that use topical information to compute topic similarity, each peer stores profiles of other peers. A neighbor profile is information that a peer maintains to describe the content stored by a neighbor. By analyzing neighbor profiles, peers try to increase the probability of choosing appropriate peers to route queries. Algorithms that use topic similarity to guide the search process in a P2P network lead to the spontaneous formation of semantic communities through local peer interactions [22].

A key concept associated with the use of topic similarity to guide the search pro-23 cess is "semantic locality" [140]. Traditionally, the notion of semantic locality has been 24 used to refer to the ability to store information about peers offering semantically close 25 services. This ability can be used to index and locate content, complementing the cur-26 rent service discovery mechanisms in a Grid and in the Web [115,150,116,79]. Semantic 27 locality has also been defined as "a logical semantic categorization of a group of peers 28 sharing common data" [119]. With the help of locality information, an unstructured P2P 29 network allows to design more informed mechanisms for routing queries, mitigating the 30 complexity of the search process [102,103]. 31

Another alternative for capturing semantic aspects consists in storing past experiences from the interaction of a peer with its neighbors. This approach does not require storing nodes' profiles. Instead, a peer keeps track of valuable routing information (such as the number of hits per node or peer availability) and uses this information to select the most active peers to forward a query [49]. The main problem with this approach is that it strongly benefits those peers that store large amounts of data, penalizing less resourceful peers that may also offer relevant material for specific topics.

A number of algorithms use heuristic information based on the state of the nodes and their past performance to select candidate nodes. There are many heuristics that can be considered; some of them are based on the analysis of the latency or the response time of specific nodes [151]. Clearly, in such cases, additional specific data must be collected and stored for computing the associated heuristic. For example, in [82] a heuristic-based query routing algorithm is presented. This algorithm collects a plurality of metrics for each host that it is aware of in a P2P network, most often by host information or query hit messages.

The metrics collected aid in determining a set of P2P hosts best able to fulfill a query 1 message, without having knowledge of specific content. The metrics collected also aid in 2 managing query messages in that they determine when to drop query messages or when to 3 resend query messages. The choice of the heuristic is a very important step in the process of developing an algorithm. In [141] a study of the *DirectedBFS* algorithm implemented 5 under different heuristics is presented. In this algorithm, in order to intelligently select neighbors, a node maintains simple statistics on its neighbors, such as the number of 7 results received through that neighbor for past queries, or the latency of the connection with that neighbor. From these statistics, the authors develop a number of heuristics to 9 select the best neighbor to send the query, such as: 10

- Select the neighbor that has returned the highest number of results for previous queries. 11
- Select the neighbor that returns response messages that have taken the lowest average 12
- number of hops. A low hop-count may suggest that this neighbor is particularly close 13 to nodes containing useful data. 14
- Select the neighbor that has forwarded the largest number of messages (all types). A 15 high message count would imply that this neighbor is stable and it can handle a large 16 flow of messages. 17
- Select the neighbor with the shortest message queue. A long message queue implies 18 that the neighbor's pipe is saturated, or that the neighbor has died. 19
- In summary, the use of semantic aspects helps to select the most promising nodes to 20 route queries with the purpose of implementing more informed search strategies. 21

#### Algorithm classification based on semantic aspects 2.3. 22

- In unstructured P2P networks, routing algorithms can be classified into content-oriented 23 or query-oriented, based on the semantic aspects and the decision-making criteria used to 24 route a query [26]. 25
- Content-oriented: these routing algorithms use metadata extracted from the shared 26 content of each peer to build a local index with global information. This index pro-27 vides each peer with an approximate view of the network content and other peers' 28 profiles. Hence, peers will be able to route efficiently their queries, improving the 29 retrieval effectiveness. Nodes' profiles are the most used semantic aspects in content-30 oriented routing algorithms [43,76]. 31
- Query-oriented: these routing algorithms exploit the historical information of past 32 queries and query hits to route future queries. Past experience based on query hits is 33 the most used semantic aspect in query-oriented routing algorithms [131,81].
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Content-oriented algorithms produce a very large number of messages to build their 35 associated indices. In contrast, query-oriented methods are more advantageous, as no ex-36 cessive network overhead is required for building the routing indices. Recently, efficient 37 approaches to content-oriented routing algorithms have been proposed in the context of 38 content-centric networking [39,146]. In content-centric networking, a data object is re-39 trieved based on its content identity instead of the IP address of the node on which it 40 resides. 41

#### 1 2.4. Distributed Hash Tables in P2P systems

Distributed Hash Tables (DHTs) are data structures for indexing data using a distributed 2 approach. DHTs provide a powerful tool that has changed the way resources and infor-3 mation are shared. In structured P2P systems, the data objects are stored by a globallyagreed scheme. In this context DHTs have turned out to be one of the most highly used 5 approaches [52]. In P2P systems implemented with DHTs, each peer represents a hash table bucket with a global hash function. Search in this kind of systems is guaranteed and efficient since it typically involves logarithmic time with respect to the overlay network size. A popular P2P system based on DHTs is Kademlia [91], which includes several de-9 sirable characteristics that were not present in previous DHT-based approaches. Kademlia 10 minimizes the number of configuration messages that every node needs to send in order 11 to learn about each other. In this system, each node has enough knowledge to be able to 12 proceed with query routing using low-latency paths. Recent work on Kademlia has been 13 oriented towards analyzing the resilience against failing nodes and communication chan-14 nels [61] and secure and trustable distribution aggregation [57]. Viceroy [90] is another 15 P2P system whose relevance lies on being the first P2P system to combine a constant 16 degree with a logarithmic diameter, while still preserving fairness and minimizing con-17 gestion. This is achieved through a quite complex architecture that guarantees with high 18 probability that the congestion of the network is within a logarithmic factor of the opti-19 mum. Later work on Viceroy resulted in Georoy [54], an algorithm for efficient retrieval 20 of information based on the Viceroy P2P algorithm. Unlike Viceroy, Georoy establishes 21 a direct mapping between the identification of a resource and the node which maintains 22 information about its location. In spite of all their advantages, it must be remarked that 23 systems based on DHTs suffer from limitations in terms of robustness and search flexibil-24 ity. A good P2P structured system needs cooperation among peers to maintain flexibility 25 and credibility. This assumption is particularly strong, as not all devices on the Internet 26 connected through the network are necessarily stable and reliable. 27

#### 28 2.5. Applications of P2P systems

<sup>29</sup> Many software applications that gained popularity among a large community of users, <sup>30</sup> such as *eMule* [14] and *PopCorn Time* [17], operate on P2P networks. In both cases, these <sup>31</sup> applications use P2P technologies to stream audio and video to their end users, generating <sup>32</sup> a considerable portion of the overall traffic on the Internet and requiring a large amount <sup>33</sup> of energy consumption [34].

Regarding to educational settings, P2P technologies have allowed institutions to share 34 files globally, as is the case of the *LionShare* project[96]. Another popular distributed ap-35 plication is *Bitcoin* and its alternatives, such as *Peercoin* and *Nxt*, which are P2P-based 36 digital cryptocurrencies. *Bitcoin* [51] is a P2P system where transactions take place di-37 rectly among users, without an intermediary. Network nodes are in charge of verifying 38 these transactions, which are eventually recorded in a public distributed ledger (called the 39 blockchain). Bitcoin has no central repository (nor administrator) and is known as the first 40 decentralized digital currency [98]. 41

Another area where P2P technologies are becoming increasingly important is social networking. Currently, the social web is mostly dominated by centralized social networks such as *Twitter*, *Facebook* and *MySpace* [18,15,16] and as a consequence it is limited by

the centralization in the use of the information and the potential loss of control of the 1 privacy of the information by the users. These social networks create the illusion that 2 users are directly connected to each other. However, the centralized servers belonging 3 to companies are the ones in charge of controlling the data and the interactions among users. The lack of control of the users on their data is a problem that is aggravated by the 5 terms of services, which are often unclear or have notable disadvantages, and by the occasional changes in the privacy policy. Distributed technologies offer a possible solution to 7 the problem of centralization. With this approach, servers (peers) can communicate with each other without the intervention of a central server. This schema allows that data, and the control over them, to be distributed among all users and their servers. P2P networks 10 offer a way to relax the constraints of centralized control, resulting in systems that are 11 decentralized, concurrent and with collective or emerging behaviors. These features make 12 P2P an attractive technology to support social networks. An example of open source dis-13 tributed social network is BuddyCloud [2], which allows software developers to share their 14 applications, supplemented with chats and videos. Another distributed social network is 15 Diaspora\* [3], whose policy allows the decentralization in the use of information. In this 16 social network profiles are stored in users' personal web servers, allowing them to have 17 full control of the content they share and to have absolute knowledge of where the con-18 tent is stored and who has access to it. Other examples of distributed social networks 19 are Friendica [5], GNU social [6], Mastodon [8], Minds [9], Kune [7] and Twister [10]. 20 Clustering is an important data mining issue, especially for large and distributed data 21

analysis. Distributed computing environments such as P2P networks involve separated 22 sources, distributed among the peers. According to unpredictable growth and dynamic 23 nature of P2P networks, data of peers are constantly changing. Due to the high utilization 24 of computing and communication resources and privacy concerns, processing of these 25 types of data should be applied in a distributed way and without central management. 26 In this scenario, clustering algorithms became important to organize the peers among 27 the network. An example of this kind of algorithm is GBDC-P2P [27]. The GBDC-P2P 28 algorithm is suitable for data clustering in unstructured P2P networks and it adapts to the 29 dynamic conditions of these networks. In the GBDC-P2P algorithm, peers perform data 30 clustering with a distributed approach only through communications with their neighbors. 31 Distributed data storage is another area in which P2P technologies have proven to be 32 helpful. Distributed databases allow quick access to data stored throughout a network, and 33 have different capabilities (e.g. some provide rich query abilities whereas others are re-34 stricted to a key-value store semantics). Google's Bigtable [12], Amazon's Dynamo [13], 35 Windows Azure Storage [19], and Apache Cassandra [11] (formerly Facebook's data 36 store [59]) are examples of distributed databases. In P2P network data storage, the user 37 can usually reciprocate and allow other users to use their computer as a storage node as 38 well. Information may or may not be accessible to other users depending on the design of 39 the network. 40

## **3.** Routing algorithms

In this section, we analyze around forty different algorithms as well as some of their
variants related to intelligent query routing in P2P networks. In particular, as discussed
previously, we will focus on unstructured P2P networks in which semantic issues play a

mayor role. We provide a brief description of each algorithm along with the corresponding
 references.

#### **3 3.1.** Routing algorithms in structured P2P networks

In a structured P2P system, the topology that defines the connections among peers and 4 data locations is predefined. This pre-established topology is exploited by search mech-5 anisms that take advantage of these pre-defined relations among peers. Even when using 6 a distributed hash table (DHT), structured systems may differ on the data structures used for implementing it (e.g. some of them may rely on flat overlay structures while others 8 might be based on hierarchical overlay structures). A benefit of DHTs is the possibility to exploit the structure of the overlay network for sending a message to all nodes, or a 10 subset of nodes, ensuring a threshold for the overall execution time involved. It is not 11 natural to implement a search algorithm with DHT in unstructured networks due to their 12 lack of structure. However, some authors have explored their application in unstructured 13 and semistructured networks [62]. 14

Some flat data structures (Figure 3) include ring, mesh, hypercube, and special graphs 15 such as the de Bruijn graph [46]. For example, Chord [124] uses a ring data structure with 16 node IDs. Each node keeps a table that contains the IP addresses of those nodes that are 17 half of the ID ring away from it. A key k is mapped to a node A whose ID is the biggest 18 that does not exceed k. In the search process, A forwards the query for key k to succ(k)19 (node in A's table with the highest ID that is not larger than k). In this way, a query can be 20 forwarded until the node that holds the key is reached. The so-called "finger table" speeds 21 up the lookup operation, ensuring an execution time of O(loqN). 22



Fig. 3. Some examples of possible structured topologies.

Pastry [114] is based on a tree data structure which can be considered a generalization of a hypercube. Each node A keeps a leaf set L. For every node A, the set L consists of Intelligent query processing in P2P networks: semantic issues and routing algorithms

those L/2 nodes whose IDs are nearest to and smaller than the ID of A, along with the set 1 of L/2 nodes whose IDs are nearest to and bigger than the ID of A. This set ensures the 2 correctness of the process. Every Pastry node also keeps a routing table of references to 3 other nodes of the same ID space. In Pastry, given a search query q associated with a key k, a node A forwards q to a node whose ID is the nearest to k among all the nodes known 5 to A. Node A tries then to find a node in its leaf set. If that node does not exist, A looks for a candidate node in its routing table whose ID shares a longer prefix with k than A. 7 If this node does not exist either, A forwards q to a node whose ID has the same shared prefix than A but is numerically closer to k than A. In this way, each Pastry node ensures 9 an execution time of O(log N). The approach presented in [149] called *Tapestry* is similar 10 to the previous one. They differ in the underlying routing algorithm and in the approach 11 taken to exploit the locality. Tapestry also ensures an execution time of O(logN). 12

In CAN [110] DHTs are implemented using a d-dimensional toroidal space divided 13 into hyper-rectangles, which define different zones. Each of these zones is controlled by 14 a particular node. Keys are mapped with a hash function to points in the d-dimensional 15 space. Each node has a routing table that consists of all of the other nodes that are in its 16 d-dimensional space. A node A is in the same space of another node B if the zone of B 17 shares a (d-1)-dimensional hyperplane with the zone of A. Given a query q associated 18 with a key k, a node forwards q to another node according to its routing table whose zone 19 is the nearest to the zone of the node responsible for key k. 20

The *NetSize-aware* protocol introduced in [148] is based on *CAN*. The main objective of this algorithm is to solve the problem of search flexibility in DHT. This algorithm preserves CAN's simplicity, providing a greedy routing algorithm based on DHT. *NetSize-aware* uses a binary partition tree algorithm to determine the underlying network topology. Simulation results show that this approach is resilient, efficient and improves the performance of *CAN*.

KaZaA [1] is a P2P file-sharing technology that was commonly used to exchange 27 MP3 music files and other file types (such as videos, applications, and documents) over 28 the Internet. Its architecture is based on a two-tier hierarchy in which some nodes are 29 distinguished as supernodes. Supernodes are those nodes with the fastest Internet con-30 nection and best CPU power. Each supernode is responsible for indexing the files of the 31 nodes that it handles. The use of supernodes with better computing capabilities than reg-32 ular nodes allows the system to perform better than the local-indices approach respect to 33 lower susceptibility to bottlenecks, and similar resilience to churn (where the churn rate 34 can be defined as a measure of the number of individuals or items moving out of a collec-35 tive group over a specific period of time). However, this system suffers the problem of the 36 resulting overhead associated with exchanging index information between regular nodes 37 and supernodes [143,84]. In [65] a routing algorithm that is also structured in two layers 38 is proposed: *SkipNet* layer and *Small-World* layer. The first layer routes the queries based 30 on a numerical ID and the second layer routes the queries using a Small-World topology 40 (see [137] for a pioneering study of Small-World networks). 41

An efficient P2P information retrieval system called *pSearch* is presented in [127]. *pSearch* supports semantic-based full text searches and avoids the scalability problem of certain systems that employ centralized indexing. In *pSearch* documents are organized based on their vector representations generated by information retrieval algorithms based on the vector space model and latent semantic indexing. This organization results in more

efficiency and accuracy, as the search space for a particular query is defined on the basis
 of related documents.

The growth of intercommunication between computers gives systems the chance to 3 operate more efficiently, by better supporting the cooperation between individual com-4 ponents. AFT [106] is an overlay that adapts to a changing number of nodes in a P2P 5 network and is resilient to faults. The AFT overlay is designed to be a solution for sys-6 tems that need to share transient information, performing a synchronization between various components, such as in mobile ad-hoc networks, urban networks, and wireless sensor 8 networks. The operations supported by the overlay, such as joining, leaving, unicast trans-9 mission, broadcast sharing and maintenance can be accomplished in time complexity of 10  $\mathcal{O}(\sqrt{N})$ , where N is the number of nodes which are part of the structure. 11

In [146] a novel framework is introduced, based on implementing a hybrid forwarding 12 mechanism. This approach allows discovering content in a proactive or reactive way based 13 on content characteristics. The proposed framework classifies time-sensitive data utilizing 14 content identifiability and content name prefixes, aiming at applying the most suitable 15 strategy to each category. For proactive content dissemination they propose a Hierarchical 16 Bloom-Filter based Routing algorithm (see [35] for a detailed review of the concept of 17 bloom filters). A Hierarchical Bloom-Filter is structured in a self-organized geographical 18 hierarchy, which makes the approach scalable to large metropolitan Vehicular Ad-Hoc 19 Networks (VANETs). 20

An approach presented in [79] characterizes the notion of semantic-based sub-spaces 21 as a basis for organizing the huge search space of large-scale networks. Each sub-space 22 consists of a set of participants that share similar interests, resulting in semantic-based 23 Virtual Organizations (VOs). Thus the search process occurs within VOs where queries 24 can be propagated to the appropriate members. The authors propose a generic ontological 25 model that guides users in determining the desired ontological properties and in choosing 26 the "right" VOs to join. DHTs are used to index and lookup the hierarchical taxonomy 27 in order to implement the ontology directory in a decentralized manner. Even though the 28 ontology-based model facilitates the formation of the VOs, searching and sharing effi-20 ciently is still a major challenge due to the dynamic and large-scale properties of the 30 search space. In order to efficiently share and discover resources inside VOs an infras-31 tructure called OntoSum is proposed. 32

Security is an important feature in all type of networks, especially in P2P networks 33 where every participant requests and provides information without any centralized control. To prevent structured overlay networks from being attacked by malicious nodes, a 35 symmetric lookup-based routing algorithm referred to as Symmetric-Chord is presented 36 in [87]. This algorithm determines the precision of routing lookups by constructing multi-37 ple paths to the destination. The selective routing algorithm is used to acquire information 38 on the neighbors of the root. The authenticity of the root is validated via consistency 39 shown between the information ascertained from the neighbors and information from the 40 yet-to-be-verified root, resulting in greater efficiency of resource lookup. Simulation re-41 sults demonstrate that Symmetric-Chord has the capability of detecting malicious nodes 42 both accurately and efficiently, so as to identify which root holds the correct key, and 43 provides an effective approach to the routing security for the P2P overlay network. 44

Another approach that implements an attack detection method is presented in [66]. The authors propose a routing table "sanitizing" approach that is independent of a speIntelligent query processing in P2P networks: semantic issues and routing algorithms

cific attack variant. The proposed method continuously detects and subsequently removes
 malicious routing information based on distributed quorum decisions, and efficiently for wards malicious information findings to other peers which allows for progressive global
 sanitizing.

In [23], the authors have proposed a scalable solution for lookup acceleration and optimization based on the de Bruijn graph with right shift. The proposed solution is principally based on the determination and elimination of the common string between source and destination. This procedure is executed locally at the current requestor node. The performance aspects of the proposed model have been validated through simulation results developing a specific Java program. Among other approaches that use de Bruijn graphs we can cite D2B [53], DH-DHT [99] and Koorde [70].

### 12 3.2. Routing algorithms in semi-structured or loosely structured P2P networks

In loosely structured P2P networks the overlay structure is not strictly specified. The 13 emerging structure turns out to be formed in a probabilistic way, or defined by some 14 underlying topology. Thus, searching in this kind of networks depends on the overlay 15 structure and how the data is stored [125]. FreeNet [41,4] is a P2P loosely structured 16 system designed for protecting the anonymity of data sources. This scheme is based on 17 the DHT interface, where each node has a local data repository and an adaptable routing 18 table. These tables have information about addresses of other nodes and the possible keys 19 stored in these nodes. Searches are performed in the following way: let us assume that 20 node A is the query-issuing node and it generates a query q for a data item with key k. 21 First, A looks up its data repository. If the file is found locally, q is resolved. Otherwise, 22 q is forwarded to the node B whose key is nearest to k according to As routing table. 23 Then, node B performs a similar computation. This procedure continues until the search 24 process terminates. During this process, a node may not forward the query to the nearest-25 key neighbor because that neighbor is down or a loop is detected. In such cases, this node 26 tries to contact the neighbor with the second nearest key. A TTL (time to live) limit is 27 specified to restrict the number of messages in the query routing process. If the data item 28 is found, the file is returned to the query-issuing node in the reverse path of the query. 29 Each node (except the last one on the query path), creates an entry in the routing table for 30 the key k. To bring anonymity, each node can change the reply message and claim itself 31 or another node as the data source. 32

*PHIRST* [113] is a system that aims at facilitating full-text search within P2P databases 33 and simultaneously takes advantage of structured and unstructured approaches. In a sim-34 ilar way to structured approaches, peers publish first terms within their document space. 35 The main difference with respect to other algorithms is that frequent terms can be quickly 36 identified and do not need to be stored exhaustively, thus reducing the storage require-37 ments of the system. In contrast, during query lookup agents use unstructured search to 38 compensate for the lack of fully published terms. In this way the costs of structured and 39 unstructured approaches are balanced, achieving a reduction in the costs involved in the 40 queries that are generated in the system. There are other kinds of semi-structured P2P net-41 works where the network is divided into different subnets, resulting in a topology based 42 on the peers' interests. In [93] a system is presented where nodes are clustered according 43 to their interests [33] to form a P2P overlay network of multilayer interest domains. Three

types of nodes are distinguished: active nodes, super-nodes, and normal nodes. Each active node acts as a router providing information that facilitates query routing information
at the cluster level. Each super-node is responsible for maintaining the related information of each member of the cluster. Finally, normal nodes are responsible for providing
and sharing resources. The network resulting topology is shown in Figure 4.

There are other kinds of semi-structured P2P networks where the network is divided into different subnets, resulting in a topology based on the peers' interests. In [93] a sys-7 tem is presented where nodes are clustered according to their interests [33] to form a P2P overlay network of multilayer interest domains. In this system, there are three types of 9 nodes: active nodes, super-nodes, and normal nodes. Each active node acts as a router 10 providing information that facilitates query routing information at the cluster level. A 11 super-node is a representative node of a cluster and is in charge of maintaining the related 12 information of each member node in the cluster. Finally, a normal node is mainly respon-13 sible for providing various types of shared resources. The resulting network topology is 14 shown in Figure 4. Another similar approach is presented in [133], where the authors pro-15 pose a system in which clusters are constructed on multiple logical layers. In this system, 16 peers can switch overlay networks to search content based on popularity. One of the over-17 lay networks is a network based on clusters constructed according to the content of each 18 peer [134,104]. 19

Social media has changed our way of communication and sharing data on the Inter-20 net, which is now mostly based on collaboration among members to provide and exchange 21 information. The efficiency of this new form of interaction motivates researchers to de-22 sign architectures based on the social behavior of the users. In [30] an algorithm called 23 *ROUTIL*, that combines social computing and P2P systems, is presented. The goal in 24 ROUTIL is to link users with similar interests in order to provide a secure and effective 25 service. A method for modeling users' interest in P2P-document-sharing systems based 26 on k-medoids clustering is presented in [108]. In the proposed approach an overlay net-27 work is created based on the k-medoids clustering algorithm, which is combined with the 28 users' historical queries to improve the initial user interest model. 29

#### **30 3.3.** Routing algorithms in unstructured P2P networks

In an unstructured P2P network (Figure 5), there is no specific criterion which strictly de-31 fines where data is stored and which nodes are neighbors of each other. The Breadth First 32 Search (BFS) or flooding is the typical algorithm used to search in pure P2P networks. In 33 these algorithms, queries are propagated from a node to all of its neighbors, then to the 34 neighbors of those nodes and so on, until the TTL parameter becomes zero. This routing 35 method is implemented in some systems such as Napster and Gnutella [100,111]. Flood-36 ing tries to find the maximum number of results. However, flooding does not scale well 37 [111], and it generates a large number of messages in comparison with other approaches. 38 Many alternative schemes have been proposed to address the original flooding prob-39 lems in unstructured P2P networks. In *iterative deepening* [142], also called expanding 40 ring, the query-issuing node periodically carries out a sequence of BFS searches with 41 increasing depth limits  $D_1 < D_2 < ... < D_n$ . The query is considered to be resolved 42 when the query result is satisfied or when the maximum depth limit *n* has been reached. 43 In the latter case, the query is assumed to remain unsolved, and it can be determined 44 that the query-issuing node will never find the answer to that query. All nodes use the 45



Fig. 4. Network topology structure in a P2P overlay network of multilayer interest domains (adapted from [93]).



Fig. 5. Graphical representation of an unstructured topology in a P2P network.

same sequence of depth limits called *policy P* and the same period of time between two
 consecutive BFS searches. This algorithm is appropriate for applications where the ini tial number of query hits is important, but this approach does not reduce the number of
 duplicate messages and the associated query processing time is high.

In a *Depth-First Search* (DFS) algorithm, rather than sending a query to all the neighbors, each peer selects a single candidate neighbor to send the query. In this scheme, the maximum TTL of a query is used to specify the search depth. If the query-originating node does not receive a reply within a certain period of time, the node selects another neighbor to send the query. The process is repeated until the query is answered or all the neighbors have been selected. The criteria used to select a neighbor can highly influence the performance of the search process. FreeNet [41,4] is an example of a P2P system using the DFS scheme.

In the standard *random-walk* algorithm [55], the query-issuing node forwards the 13 query to one neighbor selected randomly. On its turn, this neighbor proceeds in a similary 14 way, choosing randomly one of its neighbors and forwarding the query message to that 15 neighbor. The procedure is repeated until the required data is found. This algorithm uses 16 only one walker, reducing the message overhead but causing longer search delays. In the 17 k-walker random walk [88], k copies of the query message are sent by the query-issuing 18 node to k randomly selected neighbors. Each query message takes its own random walk. 19 In order to decide if a termination condition has been reached, each walker periodically 20 communicates with the query-issuing node. This algorithm attempts to reduce the routing 21 delay. A similar approach is the two-level random-walk algorithm [67]. In this algorithm, 22 the query-issuing node uses  $k_1$  random search threads with a TTL with a value of  $l_1$ . 23 When this TTL parameter expires, each search thread explodes to  $k_2$  search threads with 24 the TTL parameter established in  $l_2$ . This approach aims to reduce duplicate messages, 25 but it has a longer search delay than the k-walker random walk. Random-walk approaches 26 are popular in P2P applications. For example, in [80] a study of the random-walk dom-27 ination problem is presented with the formulation of an effective greedy algorithm that 28 guarantees an optimal performance. 29

Another similar approach is the *modified random BFS* algorithm [71] where the queryissuing node forwards the query to a randomly selected subset of its neighbors. On receiving a query message, each neighbor forwards the query to a randomly selected subset of its neighbors (excluding the query-issuing node). This algorithm continues until some stop condition is satisfied. As pointed out in [71], this approach results in more nodes being visited and has a higher query success rate than the *k*-walker random walk.

Directed BFS [142] is a routing algorithm that selects those neighbors from the query issuing node which are expected to quickly return many high-quality results. The selected
 neighbors subsequently forward the query message in a BFS way to all their neighbors.
 Each peer stores simple statistics about its neighbors (e.g. the highest number of query
 results returned previously, network latency for the neighbor, or the least busy neighbors)
 and uses this information for a more informed neighbor selection strategy.

Intelligent search [71] is similar to directed BFS. However, a more intelligent approach to neighbor selection is achieved by considering the past performance of the queryissuing node neighbors and limiting query propagation only to a selected subset of these neighbors. These neighbors are selected through a query-oriented approach that considers whether the neighbors have successfully answered similar queries (based on query cosine similarity [64]) in the past. Each node keeps a profile of its neighbors with information
 on those queries that the neighbors answered more recently in the past. Similarly to other
 query propagation approaches, a TTL value is used to stop the query propagation process.

In the local-index-based search algorithm [142], every node replicates the indices maintained by other nodes for their local data with a k-hop distance from it. In this way, 5 a node can use data from its local indices to answer queries associated with data stored 6 in other nodes. A broadcast policy P defines when query propagation must stop. As a consequence, only those nodes at depths smaller than those listed in P check their local 8 indices and return the query result if the requested data is found. Local indices are updated to reflect changes when a node joins, leaves, or modifies its own data. At the time a node 10 Y joins the network it sends a join message with a TTL of r hops. Hence, all nodes 11 within an r-hop distance from Y receive this message. The message contains metadata 12 describing Y's data collection. If a node X receives a join message from Y, it replies 13 with another join message with metadata describing its own data collection to keep Y's 14 index up to date. Each time a node Z leaves the network or dies, other nodes that index 15 Z update their indices after a timeout, removing information on Z's data collections from 16 their indices. Modifications on Z's data collections are reflected on other nodes indices by 17 sending a short update message with a TTL of r to all Z's neighbors. Query propagation 18 in the local-index-based search approach is similar to the iterative deepening approach in 19 that both algorithms rely on a list of depths to limit the number of hops allowed. However, 20 while in iterative deepening nodes maintain indices containing local information only, in 21 local-index-based search, nodes maintain indices containing not only local information 22 but also information about data collections from other nodes. 23

The routing-index-based search algorithm [43] is similar to directed BFS and intelli-24 gent search. The three approaches guide the entire search process using neighbor infor-25 mation but differ in the type of information stored and the way this information is used. 26 In directed BFS only the query-issuing node uses this information to select appropriate 27 peers while the rest of the nodes use BFS as a strategy to route queries. Intelligent search 28 uses information about the past queries that have been answered by neighbors. However, 29 different from the rest, routing-index-based search stores information about the number of 30 documents and the topics of the documents stored in the neighbor nodes. This facilitates 31 the process of selecting the best candidate neighbors to forward queries. In routing-index-32 based search, good neighbors typically provide a means to quickly find many documents. 33 Since indices are required to be small in a distributed-index mechanism routing indices do 34 not maintain the location of each document. Instead they maintain information to guide 35 the process of finding a document. 36

To illustrate the routing indices approach, we revisit the example presented in [43]. 37 Consider Figure 6 which shows four nodes A, B, C, and D, connected by solid lines. The 38 document with content x is located at node C, but the RI of node A points to neighbor B39 instead of pointing directly to C (dotted arrow). By using "routes" rather than destinations, 40 the indices are proportional to the number of neighbors, rather than to the number of 41 documents. The size of the RIs is reduced by using approximate indices, i.e., by allowing 42 RIs to give a hint (rather than a definite answer) about the location of a document. For 43 example, in the same figure, an entry in the RI of node A may cover documents with 44 contents x, y or z. A request for documents with content x will yield a correct hint, but 45 one for content y or z will not. This is a content-oriented method such that the node

knowledge about topics belonging to other peers is updated when a node establishes a
 new connection (not by past experience).



Fig. 6. Routing Indices schema (adapted from [43]).

Some P2P information retrieval methods are adaptations of a classification problem to query routing. In a classification problem, the classifier tries to classify an object using Δ some features. The Semantic Overlay Model [68] aims at locating appropriate peers to 5 answer a specific query. Instead of broadcasting queries, this approach routes queries 6 to semantically similar peers. They produce semantic vectors in order to classify peers into categories that represent the peers' semantic similarity. This is a content-oriented 8 method given that it uses meta-information to classify peers by interests. A query can be 9 routed to related peers, increasing the recall rate while reducing at the same time hops 10 and messages. Experiments have shown that establishing a semantic overlay model based 11 on latent semantic indexing and support vector machine methods is feasible and performs 12 well and that the query routing algorithm in the semantic overlay is efficient [68]. 13

In a pure P2P unstructured and decentralized network all the peers usually have the 14 same responsibilities. However, some query-routing methods in unstructured P2P network 15 make use the notion of "super-peer" as is the case in the *Backpressure* algorithm [118]. 16 In this algorithm, super-peers serve their subordinates by resolving queries or forwarding 17 them to other super-peers. Super-peers can resolve queries by checking the files/resources 18 they have, as well as those of their subordinate community. Methods that impose some 19 structure on special peers are considered "routing algorithms for semi-structured P2P net-20 works", but in this case super-peers are self-organized without a central or initial control. 21 The algorithm *Backpressure* is query-oriented, and it uses past information to decide how 22 to route queries disregarding the content of each peer. 23

The *Route Learning* algorithm [40] uses keywords extracted from queries to deter-24 mine how to route the queries. This differs from other approaches, where meta-data is 25 used to classify queries and to decide how to route them. In the Route Learning scheme, a 26 peer tries to estimate the most likely neighbors to reply to queries. Peers calculate this es-27 timation based on the knowledge that is gradually built from query and query hit messages 28 sent to and received from the neighbors. Route Learning reduces the query overhead in 29 flooding-based networks using keywords extracted from queries, being therefore a query-30 oriented method. 31

Routing future queries using past experiences is the best way to route queries to specific nodes with the objective of improving performance, but it is important to store this accumulated knowledge in an efficient way. As a consequence, each peer may need to
 consider some storage space for maintaining metadata. This storage also implies the cost
 of keeping track of these data updates. The *Learning Peer Selection* approach [26] im plements a query-oriented method on unstructured networks with the ability to discover
 users' preferences by analyzing their download history. The proposed model is imple mented in three layers. The first of these layers is especially dedicated to store and to
 update past information. The other two layers are responsible for managing users' pro files and selecting the relevant peers to send queries.

<sup>9</sup> Cooperation among peers in a P2P network is strongly linked with the concept of <sup>10</sup> knowledge sharing. Usually, there is a trade-off between improving the network global <sup>11</sup> knowledge and the cost of sending update messages through the network. The *Self Learn-*<sup>12</sup> *ing Query Routing* algorithm [38] attempts to improve global knowledge by learning the <sup>13</sup> nodes' interests based on their past search result history. The number of shared files deter-<sup>14</sup> mines a rank of friendship between two nodes. Queries are initially routed to friend nodes <sup>15</sup> only. In case of failure, a broadcast search is executed. Past search results allow nodes to <sup>16</sup> incrementally learn about other nodes in the network that share the same interests.

A P2P algorithm that relies on the notion of semantic communities is *INGA* [83]. The 17 INGA algorithm assumes that each peer plays a different role in a social network, such as 18 content provide, recommender, etc. The roles associated with peers allow INGA to deter-19 mine the best matching candidates to which a query should be forwarded. Facts are stored 20 and managed locally on each peer, constituting the *topical knowledge* of the peer. Each 21 peer maintains a personal semantic shortcut index. An evaluation of different P2P search 22 strategies based on the 6S system [139] is carried out in [22] with the purpose of showing 23 the emergence of semantic communities. The query-routing evaluated strategies include a 24 random, a greedy and a reinforcement learning algorithm. To route queries appropriately, 25 in the greedy and the reinforcement learning algorithms, each peer learns and stores pro-26 files of other peers. A neighbor profile is defined by the information that a peer maintains 27 in order to describe the contents stored by a given neighbor. By adapting the profile in-28 formation, peers try to increase the probability of choosing the appropriate neighbors for 29 their queries. Simulations demonstrate that peers can learn from their interactions to form 30 semantic communities even when the overlay network is unstructured. Another content-31 oriented search algorithm is State-based search (SBS) [138]. In this algorithm, each node 32 maintains a list with state information associated with the other nodes in the network and 33 uses this information to route queries. Searches are performed using a local fuzzy logic-34 based routing algorithm. Results reported by the authors indicate that SBS reduces the 35 response time and obtains a better load balance when compared to baseline algorithms. 36

An approach aimed at achieving low bandwidth is *Scalable Query Routing* (SQR) [76]. 37 In this algorithm, a routing table is maintained at each node that suggests the location of 38 objects in the network based on the past experience. A data structure called *Exponen*-39 tially Decaying Bloom Filter (EDBF) encodes probabilistic routing tables in a highly 40 compressed manner and allows efficient query propagation. Other content-oriented meth-41 ods seek to control the system congestion by tracking alternative routes to balance the 42 query load between peers. For instance, the method presented in [120] relies on a *Collab*-43 orative Q-Learning algorithm that learns several parameters associated with the network 44 state and performance. In [31] two algorithms are presented that are a combination of 45 other existing techniques. One algorithm is a combination of Flooding and Random Walk

while the other combines *Flooding* with *Random Walk* with Neighbors Table. The authors
 present different results obtained from simulations over an unstructured P2P network that
 showed that hybrid algorithms provide the most balanced performance regarding the av erage number of hops, average search time and the number of failures when compared to
 the basic resource discovery algorithms.

6 Other relevant search algorithms in unstructured P2P networks that are not described 7 in this article but are classified in the following section are *q-pilot* [126], *SemAnt* [95],

8 Remindin' [128], P2PSLN [152] and NeuroGrid [69].

## **9** 4. Comparative Analysis

Next we will present a comparative analysis of the major features involved in the query
 routing process. We will also discuss the advantages and disadvantages of the different
 existing approaches.

## 13 4.1. Features comparison

In this subsection, a comparative analysis of the algorithms previously described is pre-14 sented. Table 1 shows a comparison between routing algorithms in structured P2P net-15 works. In this kind of systems, the use of DHT allows ensuring a logarithmic execution 16 time. The algorithm used by the SkipNet and Small-Word scheme shows a central differ-17 ence with respect to the other algorithms presented in table 1. In Chord and Pastry, the 18 goal is to implement a DHT diffusing content randomly throughout an overlay in order 19 to obtain a uniform and load-balanced behavior, whereas in *SkipNet* the goal is to enable 20 systems to preserve useful content and path locality using the Small-World topology to 21 take advantage of shortcuts to remote nodes. 22

				Features	
		DHT	Flat	Overlay Hierarchical	- Structure
	Chord	•	٠		Ring
	Pastry	٠	٠		Tree
	Tapestry	٠	٠		Tree
	CAN	•	٠		Toroidal
Algorithm	KaZaA	•		•	2-layers
	SkipNet and Small World			•	2-layers
	pSearch	٠		•	Toroidal
	AFT		٠		Toroidal
	HBFR			•	Geographical
	OntoSum	٠		•	Tree

**Table 1.** Comparison of salient features that characterize structured P2P networks.

In unstructured P2P networks, search turns out to be a difficult, non-scalable process [75]. As a consequence, these algorithms take advantage of different semantic asIntelligent query processing in P2P networks: semantic issues and routing algorithms 21

- 1 pects in order to optimize their associated search processes. Table 2 outlines the semantic
- <sup>2</sup> aspects that are present in each of the unstructured systems described above. The first five
- <sup>3</sup> are flooding-like algorithms, and consequently they do not consider any semantic aspect.
- <sup>4</sup> In the rest of the algorithms, the goal is to strategically select candidate nodes in order to
- <sup>5</sup> reduce query propagation. To do that, some algorithms (e.g. Directed BFS) use heuristic
- 6 information, while others select the candidate nodes by past experience (query-oriented)
- <sup>7</sup> or by analyzing the profile of a node (content-oriented). Finally, there is a subset of algo-
- <sup>8</sup> rithms that use a classifier to decide which are the best candidate nodes.

**Table 2.** Semantic aspects in unstructured P2P networks.

		Semantic Aspects		
		Heuristic Information Content Oriented	Query Oriented C	lassification
	Flooding			
	Iterative Deeping			
	Random Walk			
	K-walker Random Walk			
	Two-level K-walker Random Walk	2		
	Modified Random BFS			
	Directed BFS	•	•	
	Intelligent Search		•	
	Local Indices Based Search	•	•	
	Routing Indices Based Search	•		
	Semantic Overlay Model	•		•
	Route Learning		•	•
Algorithm	Learning Peer Selection		•	
Aigorium	Self Learning Query Routing		•	
	6S - Random			
	6S - Greedy	•		
	6S - Reinforcement Learning	•		
	q-pilot			٠
	SemAnt	•		٠
	REMINDIN'		•	
	P2PSLN	•		
	NeuroGrid	•		
	INGA	•		
	SQR	•		
	SBS	•		
	Collaborative Q-Learning	• •		

There are some algorithms (such as BFS, DFS, and random approaches) that do not exploit semantic aspects and consequently they are forced to implement a less informed method for propagating queries. These features can be observed in table 3. From this table, we can appreciate that even those algorithms that account for semantic aspects have

a basic mechanism to propagate queries. These mechanisms are executed over the subset
of candidate nodes or when no candidate node exists and queries must be propagated in
an alternative way. Another feature that is present in this kind of algorithms is the TTL
parameter, which is decremented by one every time the message goes from a node to
another. By performing this process, when the value of the TTL parameter becomes zero
the search for candidates can be assumed to be over, so that the original message can be
ultimately discarded.

Features BFS DFS Random TTL Flooding • • Iterative Deeping • • Random Walk • • • K-walker Random Walk • • Two-level K-walker Random Walk • • ٠ Modified Random BFS • n/a • Directed BFS n/a • Intelligent Search • • Local Indices Based Search • • Routing Indices Based Search n/a . Semantic Overlay Model n/a n/a n/a • Route Learning n/a n/a n/a • Learning Peer Selection n/a n/a n/a • Algorithm Self Learning Query Routing • n/a 6S - Random • • • 6S - Greedy • • 6S - Reinforcement Learning • • q-pilot n/a n/a n/a • SemAnt n/a n/a n/a • REMINDIN' n/a n/a n/a n/a P2PSLN n/a n/a n/a • NeuroGrid • • INGA n/a n/a n/a n/a SQR n/a n/a n/a n/a SBS n/a n/a • Collaborative Q-Learning • •

Table 3. Basic routing algorithms in unstructured P2P networks.

Figure 7 presents a timeline that shows the evolution of the main routing algorithms P2P networks over the years. From this timeline we can see that between 2002 and 2005 there was a considerable growth of algorithms for unstructured networks, whereas in recent years research efforts have been particularly focused on hybrid and structured networks. Furthermore, it can be seen that among algorithms for searching in unstructured networks, intelligent algorithms are still a minority, being most of them based on basic
 flooding mechanisms.

As introduced in Section 2, query routing algorithms can be classified according to the 3 topology of the underlying network. Algorithms for structured networks use some data 4 structure in order to select destination peers. Most of these algorithms use data structures 5 such as trees or rings, but there are other less usual structures such as the geographical po-6 sition of the peers or toroidal. Algorithms for query routing in unstructured P2P networks 7 can be classified according to the degree of intelligence that they use to select destination 8 peers. Most of these algorithms are based on basic routing techniques using a random 9 parameter or simply based on graph paths. Intelligent algorithms use different strategies 10 for routing queries, which range from the adoption of particular classification techniques 11 to the use of different kinds of heuristics. Figure 8 shows the main features present in 12 structured/unstructured routing algorithms. 13



Fig. 7. Evolution of the main routing algorithms.

### 14 4.2. Discussion

P2P systems are distributed systems consisting of interconnected nodes that offer support to different applications such file sharing, distributed data storage, and distributed
social networks, among others. Developing reliable, robust, effective and efficient P2P
systems gives rise to research challenges such as the design of reputation systems in P2P



Fig. 8. Salient features of structured/unstructured routing algorithms .

- environments, the exploitation of the semantic organization of information, the use of
   cryptographic mechanisms for data protection, etc. Several design features commonly
   considered when developing P2P systems include:
- Replication. P2P systems rely on content replication to ensure content availability. Replication is a major challenge for structured systems such as *Chord*, where identifiers are linked to their location. In these cases alias are used to allow replication [124]. *CAN* utilizes a replica function to produce random keys to store copies at different locations [136].
- Security. The dynamic and autonomous nature of P2P systems poses several chal-9 lenges at the moment of ensuring availability, privacy, confidentiality, integrity, and 10 authenticity [25]. Security in P2P networks is a highly explored field. Several cryp-11 tography algorithms and protocols have been developed especially for P2P systems, 12 such as Self-Certifying Data and Information Dispersal [37,109]. Security also in-13 volves detecting and managing malicious nodes that can corrupt messages that are 14 propagated among other nodes. The "Sybil Attack" [47] is a security threat related 15 to authenticity where a node in a network claims multiple identities. The Symmet-16 ric Chord routing algorithm addresses the authenticity problem by implementing an 17 authenticity validation process. Other approaches to address security issues in P2P 18 networks rely on the use of special forms of access control lists [45]. 19
- Anonymity. Author anonymity or peer anonymity is some times required in P2P applications. An approach named "Disassociation of Content Source and Requestor"

25

adopted by *FreeNet* provides anonymity to users by preventing other nodes from 1 discovering the true origin of a file in the network. Another anonymity mechanism is 2 "Censorship Resistant Lookup", which is used by Achord [60], a variant of the Chord 3 lookup service. л Incentive mechanisms. The performance of a decentralized P2P system relies on 5 the voluntary participation of its users. To achieve a good performance it is necessary to implement methods that provide incentives to stimulate cooperation among 7 users [58]. A simple incentive mechanism is based on ranking highly the results of Ω a particular node if it has contributed significantly in previous searches. This simple method typically works for both the Web and P2P networks since appearing high up 10 in a ranking typically represents an incentive for companies and people [42]. 11 Semantic grouping of information. The semantic organization of content through 12 the emergence of semantic communities is deeply analyzed in [22]. Some systems 13 are based on the notion of "peer communities", where relationships among peers are based on nodes' interests [73]. The emergence of these communities tends to make 15

the search process more effective as it is possible to target search queries to those
communities more closely related to the topic of the queries. This feature is exploited
by some routing algorithms as discussed in [22,83].

There are several systems that use P2P technologies, such as those presented in sec-19 tion 2.5. These systems adopted different architectures and routing algorithms. We con-20 sider that no architecture is better than another, but can be use in deferents contexts. While 21 a structured architecture can guarantee a determined execution time, a decentralized one 22 can adapt more easily to topology changes. Decentralized approaches need to store some 23 data to decide how to route a query but most of the algorithms for structured topologies 24 need to keep their DHT update. Finally, as P2P technologies are still evolving, there are 25 some open research problems such as a) developing routing algorithms for maximizing 26 performance; b) defining more efficient security, anonymity, and censorship resistance 27 schemes; c) exploiting semantic grouping of information in P2P networks; d) developing 28 more effective incentive mechanisms. 29

## 30 5. Conclusions

The early Internet was designed on principles of cooperation and good engineering. In 31 many ways, it shared principles and concepts with pure P2P networks. In this decentral-32 ized scenario, algorithms that performed searches were essential. When Internet became 33 more rigid, structured and semi-structured search algorithms emerged, where collabora-34 tion among peers was no longer an important issue. Nevertheless, in the last few years pure 35 P2P networks have come back for playing a major role in the deployment of new systems 36 and technologies. Research in decentralized search has given rise to novel algorithms that 37 incorporate semantic aspects derived from the profile of each participant. These seman-38 tic aspects can be conveniently exploited to improve routing algorithms, with the goal of 39 minimizing network traffic and optimizing query response time. 40

In this article, we have reviewed the most important query routing algorithms in P2P networks, contrasting their advantages and disadvantages. To facilitate the analysis of these algorithms, we have introduced different schemes and classifications. In particular, we have discussed diverse search strategies in structured, semi-structured, and un-

structured P2P networks. Finally, we have identified common features in these networks,
 carrying out a comparative analysis for contrasting these features.

As discussed in this article, semantic issues in intelligent query routing provide a significant added value for improving search in distributed environments. This survey aims at offering an in-depth analysis of the state of the art in this exciting research area, oriented towards a wide and heterogeneous audience of researchers and practitioners working on P2P networks. New recent advances in Artificial Intelligence techniques (e.g. [103]) show that future developments in P2P networks will allow to go beyond the traditional semantic analysis by adding qualitative reasoning capabilities to the nodes. Even though some motivating preliminary results have been obtained, most of the research work in this direction is still to be done.

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Intelligent query processing in P2P networks: semantic issues and routing algorithms

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