

## Chapter 12 **Towards Collaborative Forensics**

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Mike Mabey and Gail-Joon Ahn

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**Abstract** Digital forensic analysis techniques have been significantly improved and 4 evolved in past decade but we still face a lack of effective forensic analysis tools to 5 tackle diverse incidents caused by emerging technologies and the advances in cyber 6 crime. In this paper, we propose a comprehensive framework to address the efficacious deficiencies of current practices in digital forensics. Our framework, called 8 Collaborative Forensic Framework (CUFF), provides scalable forensic services for 9 practitioners who are from different organizations and have diverse forensic skills. 10 In other words, our framework helps forensic practitioners collaborate with each 11 other, instead of learning and struggling with new forensic techniques. In addition, 12 we describe fundamental building blocks for our framework and corresponding 13 system requirements.

Introduction 15

Computer crime has swiftly evolved into organized, and in some cases state 16 sponsored, cyber warfare. The tools digital forensic examiners currently use are too 17 limited to take on the challenges that are rapidly approaching their forensic cases. 18 Before long, fundamental changes in the industry will make many of the forensic 19 techniques used today obsolete [15]. Although many contributing elements can be 20 identified, the heart of the problem is that current digital forensic examinations are 21

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M. Mabey · G.-J. Ahn (⊠)

Laboratory of Security Engineering for Future Computing (SEFCOM), Arizona State University, Tempe, AZ 85281, USA

e-mail: mmabey@asu.edu; gahn@asu.edu

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too time-inefficient. The three principal causes of this inefficiency are summarized 22 as follows:

Software Limitations: Single workstation computers have served as the primary 24 tool of our society's computing needs for a long time. With the evidence data sets 25 being as large as they are, a single computer simply does not have the resources 26 to deliver sophisticated analysis results in a timely manner.

Size of Evidence Data: Today a 1 TB hard drive can be purchased for about 28 US\$60 and the average hard drive cost per GB is less than US\$0.10 [6]. 29 Such low cost makes terabyte-sized systems commonplace among even non- 30 tech-savvy consumers. With such a proliferation of huge storage systems filled 31 with user data, examiners are confronted with a mountain of stored data to 32 work through [31]. The problem is compounded when the situation involves a 33 redundant array of independent disks (RAID) [34] or network attached storage 34 (NAS) unit shared among individuals or employees.

Increased Examiner Workload: As if insufficient tools and large datasets were 36 not enough, digital crime continues to increase in popularity [17, 24, 25], nat- 37 urally resulting in more investigations. Furthermore, state-sponsored cyberwar 38 promotes the development of increasingly sophisticated software. Simply trying 39 to keep up with the latest methods of penetration, exfiltration, and attack is 40 insufficient to accommodate the pace of digital crime.

In addition, when cases become backlogged, only those designated as more 42 urgent are worked on, potentially leaving suspects' co-conspirators at large and 43 capable of making more victims out of innocent people.

**Motivation** 45

The challenges above can be greatly reduced by a secure and robust infrastructure 46 that facilitates collaborative forensics [18, 27], which we define as the willful 47 cooperation between two or more forensic examiners during any step in the forensics 48 process, for the benefit of sharing specialized knowledge, insight, experience, or 49 tools. By this we mean to indicate a process and system through which multiple 50 examiners perform their work, using a common interface that provides the means for 51 carrying out all steps of the examination process as well as providing mechanisms 52 for collaboration between the users.

Two advantages of collaboration are of particular interest to us. First, collabora- 54 tion allows people to draw from others' expertise, which is invaluable when working 55 on problems of a diverse nature or when the problem set of a job constantly changes. 56 Second, collaboration is a method of spreading a workload, which results in less 57 time needed for the job to be completed.

Consider the following hypothetical scenario which illustrates a need for a better 59 method of collaboration. While investigating a case with multiple computational and 60 storage devices that are uncommon, Bob, who is the lead examiner, determines that 61

by soliciting the aid of two subject matter experts that he trusts, the devices could 62 be successfully examined for evidence of interest. However, when Bob makes the 63 request to his supervisor to obtain assistance from these experts, she informs him 64 that the compensation expenses of the experts' consultation fees plus travel costs 65 is too great to justify with the current budget. Bob must find a way to either make 66 do with only one of the experts or to eliminate the travel expenses. Bob needs an 67 effective means of collaborating with these experts remotely.

Similar to the above scenario, it is already quite common for evidence seizure 69 to yield a variety of digital evidence, such as a mix of Windows, Linux, and 70 Mac computers, as well as cell phones, GPS devices, gaming consoles, etc. Since 71 examiners must be certified to work on a particular type of evidence (depending 72 on the investigating agency), such a workload must be split up among personnel. 73 Since there is no tool which can accommodate all evidence types, the evidence 74 presentation lacks uniformity in format and structure.

While many generic collaboration solutions exist today, none of them have 76 been crafted specifically for the needs of the digital forensics industry. To be truly 77 effective, a collaborative forensics infrastructure should maintain the strict privacy 78 and integrity principles the discipline demands, while also giving examiners the 79 flexibility to communicate however is best for the situation. This demands a level of 80 robustness that is simply not offered by collaboration tools at present.

Beyond just communication, collaboration also implies a sharing of resources. 82 For a proper exchange of data (whether it be files needing to be analyzed or the 83 results of an analysis), there must first be a uniform representation of that data, 84 and then a common storage space solution where all collaborators can keep their 85 resources secure. This will require the establishment of standards to ensure that all 86 parties can access and interpret the data. Means to efficiently manage resources will 87 also be needed.

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If examiners are to collaborate on a large scale, it will also be crucial for 89 this infrastructure to provide vast amounts of computing power, which is best 90 accomplished through some distributed processing method. Ideally, a distributed 91 processing solution would also include scalable resources. Because there is not a 92 single technological solution that will properly meet this need for all organizations, 93 there must be a generic way to interface for such processing resources.

To best facilitate collaboration among examiners, a collaborative forensics 95 solution should not be limited to supporting its use on a small number of operating 96 systems. This would hinder the collaboration process and may exclude experts who could offer potentially crucial insight.

The rest of this paper is organized as follows. We first discuss the progress 99 made by others in related fields in section "Related Work". In section "CUFF: 100 Collaborative Forensic Framework" we provide the architecture of our solution, 101 which is an abstraction of the most essential components. We then introduce all 102 other necessary components and provide details on how to realize our framework 103 in section "Realization of CUFF". Section "Conclusion" concludes this paper by 104 summarizing our contributions and discussing our future work.

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**Related Work** 106

The nature of our work is such that it brings together aspects of other, previously 107 completed works which we discuss here by topic.

**General Digital Forensics:** Two challenging types of evidence that forensics 109 examiners need to be able to analyze at times are Redundant Array of Independent 110 Disks (RAID) storage systems and drives protected with encryption. In [34], Urias 111 and Liebrock attempted to use a parallel analysis system on RAID storage systems, 112 and documented the issues and challenges they faced with that approach. Similarly, 113 multiple methods of properly handling the challenges presented by encrypted drives 114 have been presented by Casey and Stellatos in [5] and by Altheide et al. in [2].

Distributed Processing for Forensics: With distributed processing in use so 116 much today and in so many distinct settings, it is natural to think of using it to 117 divide the workload of digital forensics processing. Several years ago, when the 118 use of distributed processing was not yet as common as it is today, Roussev and 119 Richard proposed a method for moving away from single workstation processing for 120 forensic examination to a distributed environment [31]. A few years later, Liebrock 121 et al. proposed improvements upon Roussev and Richard's system in [21], which 122 introduced a decoupled front-end to a parallel analysis machine.

In [32], Scanlon and Kechadi introduced a method for remotely acquiring 124 forensic copies of suspect evidence which transfers the contents of a drive over a 125 secure Internet connection to a central evidence server. While this effort is a step 126 for the better in terms of making evidence centrally accessible, it is difficult to see 127 the direct utility of such an approach without accompanying software or analysis 128 techniques to take advantage of storing the evidence on a server. Furthermore, 129 the presented approach relies on either using the suspect's Internet connection to 130 upload the image or images, or the use of a mobile broadband connection. Given 131 the relatively abysmal upload speeds for current mobile broadband when dealing 132 with data sets that are hundreds of gigabytes or even a few terabytes large, this 133 approach will continue to be prohibitively inadequate.

Forensics Standardizations: Garfinkel has made great efforts to create stan- 135 dards to improve the overall digital forensic examination process. Garfinkel et al. 136 presented the details of a forensic corpora in [16] with the purpose of giving 137 researchers a systematic way to measure and test their tools. Garfinkel took this 138 a step further in [13] with his work to represent file system metadata with XML. 139 Finally, in [15] Garfinkel put forth a challenge to researchers and developers 140 everywhere to take note of the current industry trends and take them head on with 141 innovative forensic solutions that match the properties of emerging technologies.

Storage: Since our realization of our framework is built upon a cloud, we also 143 consider work done by researchers to address some of the issues related to shared 144 storage in a cloud. Du et al. proposed an availability prediction scheme for sharable 145 objects, such as data files or software components, for multi-tenanted systems 146



in [8]. In [36], Wang et al. introduced a middleware solution to improve shared 147 IO performance with Amazon Web Services [3]. Increasing the security of the data 148 stored in a cloud has been improved upon by Liu et al. in [22] and by Zhao et al. 149 in [38].

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In addition to the above subject areas, there also appears to be a trend toward 151 supporting collaboration mechanisms in digital forensics tools such as FTK 3 [11]. 152 But, to the best of our knowledge, there has yet to be a single system which can 153 satisfy all the functionalities set forth in section "Introduction" in a truly robust 154 manner.

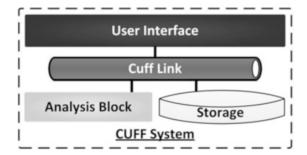
## **CUFF: Collaborative Forensic Framework**

Based on the features and requirements necessary to achieve collaborative forensics 157 as enumerated in section "Introduction" and the work presented in [23], this section 158 describes our framework, called Collaborative Forensic Framework (CUFF), and 159 elaborates what mechanisms are needed to facilitate these features. As illustrated 160 in Fig. 12.1, our framework consists of four core components (i) to mediate 161 communication between components in the system, (ii) to coordinate the distributed 162 analysis processing, (iii) to maintain the shared storage space, and (iv) to provide a 163 basic interface to the system for the user interface. While a precise set of APIs for 164 these four components may vary for the deployment setting, they should always 165 fulfill specific foundational operations and always have the same basic interactions 166 with the other components. We now discuss these two points in context of each 167 component:

Analysis Block: This component is the workhorse of the system, and in truth 169 all other components are simply in place to either provide an interface to it, or 170 to facilitate its proper function. The Analysis Block is composed of a controller 171 as well as all processing resources. Ideally, the processing resources would be 172 quite substantial and capable of handling a continuous inflow of analysis jobs of 173 significant size. The controller will receive a large number of analysis requests, 174 and is expected to enqueue and dequeue each job request in an organized and 175 efficient manner, which should also be fault-tolerant and maintain a high level 176 of responsiveness. Because it is in charge of maintaining the queue of jobs, the 177 controller oversees the processing resources and ensures that they are used properly 178 according to a selected method of prioritizing the jobs.

Storage: This component keeps track of all acquired disk images, the analyses 180 of their contents, comments and notes from users, and related information all need 181 to be kept for performing forensic tasks. To do this, it must accept incoming data 182 streams of acquired disk images, and strictly maintain the integrity of the data 183 through validation of the original checksums. Requests for getting and putting 184 data to and from the storage component will come at a high rate, particularly 185

Fig. 12.1 Each of the components in CUFF fulfill one of the four main objectives of the system. All inter-system communication passes through the Cuff Link, and the end user only interacts with the user interface component



from the processing resources in the Analysis Block, so the storage component's 186 response time needs to be controlled. To maximize reusability, a generic method 187 of transferring data to and from the storage unit should be used such that distinct 188 data types (such as analysis results, user comments, and communications between 189 users) will not need any specialization made to the system. In coordination with whatever access control mechanism is implemented, the storage component also 191 maintains strict confidentiality of the data it stores. The storage component must 192 also be flexible enough to allow temporary and/or limited access to case data for 193 subject matter experts conducting consultation work, allowing them to collaborate 194 with those directly responsible for the case.

User Interface: This is the access portal through which all the system's 196 features are made available. More specifically, the user interface supports evidence 197 acquisition, allows users to view the structure and contents of files, accepts requests 198 for specific analyses to be performed on files or groups of files, and provides a means 199 for users to communicate and share data and information with each other.

Cuff Link: This component mediates communication between all other com- 201 ponents in the system. It validates parameter input and stores location information 202 for each of the other components. Also, since it is the component that manages 203 the forensic process, it is responsible for assigning examiners jobs and notifying 204 supervisors when the work on a case has been completed. The Cuff Link maintains 205 order in the system by dictating the available APIs for each of the other components. 206 It also simplifies the implementation of other components by reducing the number 207 of connections they must make down to one.

### Realization of CUFF

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In this section, we describe how to realize the CUFF framework using commercial 210 off-the-shelf (COTS) software and open source tools. For each component, we 211 address the desired functionality, some of the challenges associated with achieving 212 such functionality, and what tools and software meet these challenges and why.

It is highly desirable for an implementation of CUFF to be easily accessible by 214 the users that will collaborate through it, to have scalable resources, and to have 215

built-in redundancy for fault-tolerance. While it would be possible to implement 216 CUFF as a set of desktop applications that communicate with other remote instal- 217 lations through a peer-to-peer networking architecture, such an approach would be 218 difficult to monitor and assure that all connected parties strictly abide by the rules 219 of evidence.

To achieve the system-wide features we desire, we have selected to build 221 upon a cloud-based infrastructure, deploying each of the components as virtual 222 machine (VM) instances. Using a cloud architecture has many obvious advantages. 223 One advantage is that VM instances can be spawned quite easily, improving the 224 scalability as well as the reliability and recovery time of all the components of the 225 system. A second advantage is derived from the fact that each of CUFF's features 226 are made accessible through various web services, including a web interface that can 227 be accessed and used by all authenticated users. Although most of the web services 228 in the system are only accessible internally, the use of web standards increases the 229 composability of the system.

While several cloud architectures exist, OpenStack [29] stands out as one built 231 for a high level of flexibility and scalability while also exporting an API compatible 232 with Amazon EC2 and S3 services [3], hence allowing the use of the widely-used 233 euca2ools [10] set of cloud administration tools.

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We now elaborate on our implementation of each of CUFF's components. 235 Although much of this section is dedicated to discussing a messaging protocol 236 (section "Scheduling Analysis Jobs"), we would propose that efficient collaboration 237 among forensic examiners depends heavily on the intelligent appropriation of the 238 analysis resources, which begins with the scheduling of their use. Hence, this is 239 a core component to address properly as we work towards our goal of facilitating 240 collaboration.

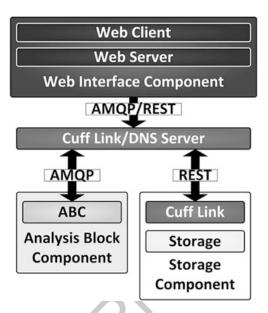
Cuff Link 242

As stated earlier, the Cuff Link provides a couple of key functions for the overall 243 system. It mediates communication, manages the forensics examination flow, 244 validates input, and exposes an API for the other components.

One thing that must be taken into consideration is that not all communication 246 types used in CUFF have the same behavior. Some types of Internet traffic are 247 difficult to stop and process before sending it on to its intended destination, such 248 as uploading or downloading files. However, with other types of traffic there is no 249 difficulty in intercepting, processing, and then forwarding the messages being sent, 250 such as requests for certain evidence files to be analyzed.

With this in mind, it would not be wise to impose a single method of handling 252 communication. Rather, the Cuff Link uses multiple technologies appropriate for 253 the type of messages being handled. Because of this, it was necessary to divide 254 the functionality of the Cuff Link such that it acts as a layer of abstraction in 255

Fig. 12.2 Inter-component communication passes through the Cuff Link, which is deployed in multiple locations to provide the layer of abstraction necessary for the various components



multiple locations. This is illustrated in Fig. 12.2 where an element of the Cuff Link 256 is running on the Storage component.

To accommodate the communication types that cannot be interrupted (which is 258 limited to traffic to and from the Storage component), the Cuff Link first provides 259 a domain resolution through Domain Name System (DNS) server, which translates 260 URLs into IP addresses. This makes it easy for other CUFF components to send 261 traffic to a destination without knowing its exact location. The component making 262 a request only needs to specify the generic name for the destination server, such as 263 http://cuff.storage.example. The Cuff Link DNS server can then resolve the name to the appropriate server.

Typically, this kind of action would require that the URL http://cuff.storage. 266 example be registered with some authoritative entity that stores all official URLs. 267 However, because we have configured the Cuff Link as a primary master name 268 server for the domains used within CUFF and specified that the system's compo- 269 nents should query the Cuff Link before any other servers, such URLs are resolved 270 within the system without making an external DNS query. This does require that 271 any CUFF components that need to be accessed by this means have their IP address 272 associated with their appropriate domain by the Cuff Link. Once this has been 273 done, however, the DNS server can also potentially perform some level of load 274 balancing among available servers by rotating which server's IP address it uses as 275 the resolution of the URL.

The second thing the Cuff Link does to accommodate communication with the 277 Storage component is to leverage a Representational State Transfer (REST) web 278 service on the Storage component. REST web services allow for certain actions 279 (in this case uploading and downloading files) to be specified in the URL of the web 280 request, using the type of web request (POST, GET, DELETE, etc.) as one factor for 281 interpreting what action should be taken. For example, a GET request in the form 282

http://cuff.storage.example/listing/2398-56-1-9125 is interpreted by first removing 283 the base URL, leaving listing/2398-56-1-9125, which is a request for the evidence 284 listing for the case number 2398-56-1-9125.

Each different type of web request is interpreted in a specific way. Within 286 the logic of these web services, we are able to perform the input validation and 287 mediation necessary for other components to access storage resources. Additionally, 288 by adding the appropriate filters, we can manage the forensics examination flow by 289 noting when certain events take place and taking action. For example, when a new 290 device image is being uploaded, the Cuff Link can easily recognize this event and 291 perform a predefined action, such as notify the appropriate supervisor that the new case needs to be assigned to an examiner.

The communication in CUFF that is of the type that can be easily interrupted 294 before reaching its destination is typically being sent to or from the Analysis Block 295 (traffic of this type are discussed in more detail in section "Scheduling Analysis 296 Jobs"). As messages are sent to the Analysis Block, they are first sent to the 297 Cuff Link and checked to ensure that the analysis request is well-formed and that 298 the specified Analysis Block is within a reachable domain. This would be the 299 mechanism whereby multiple deployments of CUFF could share analysis resources. 300

**Storage** 301

Similar to the communication in the system, there are two types of storage needs in 302 CUFF. First, because it is built on a cloud, there is a need for some way to store the VM images that run in the system. This storage need is distinct because VM images 304 are large, rarely change, but also may be needed to start up an instance very quickly. 305

Second, because cloud instances cannot store any persistent data within the 306 image itself, all data must be stored in a container suited for the particular purpose 307 of being temporarily attached to an instance and storing any data that needs to 308 be preserved. Examples of this type of data includes evidence images, evidence 309 analysis results, and database files. Evidence images will need to always be accessed 310 in a read-only mode to preserve their integrity. Furthermore, the rules of evidence 311 dictate that the system have a means of conducting logging and auditing on the 312 access of any stored data in the system.

To accommodate these features, we take advantage of the two storage facilities 314 available from the OpenStack architecture. As shown in Fig. 12.3, these facilities are 315 distinct, but together they satisfy the needs of our framework. The first is Swift, an 316 object storage component that, when used in connection with Glance OpenStack's 317 image service, can provide discovery, registration, and delivery services for virtual 318 disk images through a REST web interface. As such, Glance will act as the image 319 registry for the system.

The second storage facility we use is volumes, which are similar in functionality 321 to Amazon's Elastic Block Storage [3]. Each volume is labelled with a universally 322

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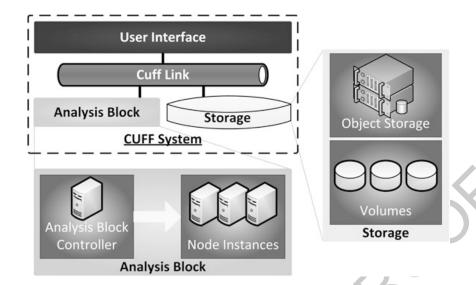


Fig. 12.3 The Analysis Block and Storage components are composed of multiple parts, each filling a separate role

unique identifier (UUID)<sup>1</sup> that distinguishes it from other volumes. The primary use 323 of these volumes will be to store evidence images and analysis results. Typically, 324 a separate volume will be created for each device uploaded to the system. If a 325 hard drive is seized that contains multiple partitions, each partition may also be 326 stored on a separate volume, to be determined by the examiner when uploading the 327 evidence to the system. Analysis results will be stored on volumes separate from 328 the evidence to which they pertain, but will store the UUID of the corresponding 329 evidence volume to keep the two connected.

In order to maintain the integrity of evidence images stored in volumes, a 331 snapshot is taken of each volume immediately after the upload is complete. This 332 essentially makes the original volume read-only because, although changes are 333 technically allowed to the volume, all write operations are saved in a "child volume" 334 that is separate from the original and can be easily discarded when the volume is no 335 longer being accessed by an instance.

One challenge that arises from doing automated, distributed analysis is finding 337 an efficient means of referring to and transferring portions of evidence images (e.g. 338 files or file segments). This is inherently a storage issue, because it is the Storage 339 component that provides access to this data. And while it is true that because we 340 have made snapshots of the evidence image volumes, we can technically attach 341 the root volume to multiple instances in a read-only fashion, this still requires that 342

<sup>&</sup>lt;sup>1</sup>UUIDs are 128-bit numbers that are used in distributed systems to uniquely identify information. The assurance that a UUID is in fact unique is derived from the number of theoretically possible numbers, which is about  $3 \times 10^{38}$ . Because of this, UUIDs are used to identify the volumes in the system.

# Author's Proof

#### 12 Towards Collaborative Forensics

```
29
          <fileobject>
            <filename>README.txt</filename>
            <id>2</id>
            <filesize>43</filesize>
            <partition>1</partition>
34
            <alloc>1</alloc>
            <used>1</used>
            <inode>6</inode>
36
            <type>1</type>
            <mode>511</mode>
            <nlink>1</nlink>
            <uid>0</uid>
40
            <gid>0</gid>
41
42
            <mtime>1258916904</mtime>
43
            <atime>1258876800</atime>
44
            <crtime>1258916900</crtime>
45
            <br/>byte runs>
46
             <run file_offset='0' fs_offset='37376' img_offset='37888' len='43'/>
47
            </byte runs>
48
            <hashdigest type='md5'>2bbe5c3b554b14ff710a0a2e77ce8c4d</hashdigest>
49
            <hashdigest type='sha1'>b3ccdbe2db1c568e817c25bf516e3bf976a1dea6</hashdigest>
          </fileobject>
```

Fig. 12.4 This portion of a DFXML file shows how a single file object is stored with all its metadata. In the case where a file is fragmented in the disk image, multiple <run> tags will be contained under the <byte\_runs> section

the instance have access to the entire image. An approach that allows for concise 343 data transfer and thorough data representation would be better.

Garfinkel's Digital Forensics XML (DFXML) representation for file system 345 metadata [13, 14] is just such an approach. We employ DFXML in CUFF to aid 346 in improving the efficiency and standardization of how CUFF stores and transmits 347 data. DFXML provides a standard way of representing and accessing the contents 348 of an imaged drive by using an XML file to store the offsets and lengths of all "byte 349 runs" (file fragments) on the disk, thereby acting as an index for the image. Using 350 this DFXML file, an entity can access specific files by simply specifying the byte 351 runs of the file and concatenating the returned results. An example of how a file's 352 information is stored in a DFXML file is shown in Fig. 12.4.

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One issue we discovered in working with DFXML is that Garfinkel's tool [12] 354 for creating DFXML files from acquired disk images was that the tool currently 355 only produces a simple Document Type Definition (DTD) specification for each 356 DFXML document, which doesn't allow for type validation. To help encourage the 357 adoption of DFXML as a standard, we have created an XML schema detailing tag 358 hierarchy and complex data types. Using this schema to validate an image's file 359 system representation, any digital forensic tool can reliably use this standard in its 360 interactions with the disk image. We believe such a schema will help developers of 361 forensic tools to be more willing to adopt this data format as a standard, because 362 they have the assurance of precise data types with any DFXML file.

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Analysis Block 364

As stated previously and as depicted in Fig. 12.3, there are two types of components 365 that make up the Analysis Block, the Controller and the Nodes. The Analysis Nodes 366 act as a network of VM instances that can be used in two different ways. First, 367 nodes can be used as a means to perform distributed, automated analysis. In this 368 case, a node is sent files to be analyzed, and the tool is executed from the command 369 line, performing the requested analysis in an automated fashion. Because analysis 370 node images are virtual machines, they can be configured to run a wide selection of 371 operating systems, maximizing platform compatibility.

The other way of using the nodes is to manually interact with them and exercise 373 fine-grained control over the work performed. A node is sent the files and the 374 specified program begins execution, but the node does not indicate to the program 375 to begin analyzing the files. Instead, the node enters a waiting state until the user 376 accesses it remotely from the system's dashboard. At this point, the user is in 377 complete control of the node and can perform whatever functions are necessary.

Since it is required that all inter-component communication in the system go 379 through the Cuff Link, communication is to be standardized and regulated through 380 the use of agents which run on all nodes in the system. While all node agents will be 381 programmed with a standard set of communication protocols, each distinct analysis 382 node's agent will be customized to the analysis programs being hosted on that 383 image. This allows the agent to store whatever parameters necessary to interface 384 with the programs as well as retrieve the analysis results. Because much of the 385 implementation for these nodes will be the same for all node types, this improves the 386 ability to flexibly support new file systems, operating systems, analysis algorithms, 387 and so forth.

**Forensic Flow** 389

The most important feature of the entire system is the fact that it accommodates the 390 main tasks of any digital forensic investigation. Since most of these tasks involve the Analysis Block in some way, we discuss the connection of each task to the Analysis 392 Block.

- 1. Acquisition: When a user is uploading a new evidence image to the system, 394 the destination volume for that image is mounted to a VM instance crafted 395 specifically for handling this task. During the upload, the instance generates 396 checksums of the image, which are validated against the checksums of the 397 original device, and then stored for later use during evidence validation after 398 operations are performed on the image. Finally, the instance takes a snapshot of 399 the volume to effectively seal it from further changes.
- 2. Validation: In order to guarantee the integrity of images and files in the system, 401 Analysis Nodes are utilized to calculate and validate the checksums before and 402 after every data transmission.



3. Discrimination: By creating a VM image that can use sets of checksums of 404 known good files (such as the Reference Data Set provided by the National 405 Institute of Standards and Technology [28]), the system can highlight those files 406 which are unknown for the examiner, effectively eliminating an extensive number 407 of files they need to look at.

To support such an event-based forensic workflow and accommodate relevant 409 workflow management features in CUFF, we consider existing approaches on a 410 web-based workflow systems [7, 20, 26]. Especially, we believe workflow modeling 411 approaches [9, 33] would help design and govern forensic flow and related tasks 412 in CUFF. In addition, BPEL (business process execution language) would be 413 another candidate to articulate a particular forensic flow for facilitating web-based 414 events [19].

## **Scheduling Analysis Jobs**

One of the merits of the Analysis Block's design is that it not only provides 417 a collection of resources for analyzing evidence, but also does so in a very 418 generic fashion, making it very reusable. The Analysis Block Controller (ABC) is 419 indifferent to both the nature of the analyses to be performed as well as what the 420 requirements are for the operating system or software used to perform the work.

While constructing a generic controller, we realized that the system had a great 422 need for scheduling automated analysis jobs, so we created a set of utilities suited to 423 accommodate this need. Our scheduling utilities have three specific purposes. The 424 first purpose is to package information regarding analysis jobs for the entities in the 425 system that will carry out the work. The information packaged includes the type of 426 analysis to be carried out, the location of the subject of the analysis (i.e. the files 427 to be analyzed), and some sort of ordering or priority information for the analysis 428 subject.

The second purpose of the utilities is to create a channel by which the information 430 described above can travel from the user who inputs it to its end destination, meaning 431 the Analysis Node that will perform the analysis specified in the job information. 432 Creating this channel implies that there is a path defined for the job that passes 433 through multiple components of the CUFF system.

The third and final purpose of the utilities is to queue jobs according to the 435 ordering or priority specified in the job information, and then to distribute jobs to 436 available nodes upon receipt of an assignment request.

A few significant properties of these objectives emerge upon examination that 438 should be highlighted. As job information is packaged, a data encapsulation method 439 must be chosen that is standard and efficient. The efficiency of both passing the 440 data along the defined path through the system and interpreting it is important 441 for preventing the utilities from becoming a bottleneck, especially since there is 442 a one-to-many relationship between the Analysis Block Controller and the Analysis 443 Nodes. No upper bound is imposed on the number of nodes in the system, nor is 444

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there any restriction on the configurations of the nodes, which again emphasizes 445 the importance of using standard data formats. Since the nodes submit requests for 446 job assignments, the scheduling utilities do not preempt the work of nodes. Finally, 447 as the utilities distribute jobs, care must be taken to prevent starvation of jobs with 448 lower ordering or prioritization.

Multiple approaches for ordering or prioritizing can be adopted depending on 450 what is most important for the deployment domain. One approach would be to 451 calculate the resource demands of each job on the system using the order of 452 complexity for each of the available analysis tools, an estimated turnaround time 453 for a small baseline job, and the size proportion difference between the baseline 454 job and that of a queued job request. With these values, CUFF could prioritize jobs 455 so that those jobs with a significant workload on the system are either spread out 456 among jobs with a lesser demand or are delayed until off-peak hours.

A second approach would be to assign priority based on the importance of the 458 case of which it is a part. At times, one investigation will be more pressing than 459 all others currently being worked on. In such a scenario, the case that has a higher 460 level of criticality should be given priority over other analysis work being done in 461 the system. By giving the user the ability to specify what criticality level a certain 462 case has been given, jobs related to that case can be allotted more time for being 463 analyzed by the system resources.

In our deployment of CUFF, we have implemented the second prioritization 465 approach. We reiterate that this scheduling scheme is specifically for automated 466 analysis. In other words, it is purposed for when users submit batches of requests to 467 analyze data segments which will then be carried out without any further input from 468 users. We anticipate examiners will interact more closely with VM instances in the 469 cloud on occasion, but that is a use case distinct from the one we address here.

To begin describing the behavior of the utilities, we first identify how we have 471 satisfied the purposes of the utilities as set forth earlier. First, the container format 472 used for all messages is JavaScript Object Notation (JSON), which is both easily 473 transmitted and easily interpreted from this notation to programming objects and 474 vice versa. Second, to satisfy the needs for a path through the system components, 475 a queuing mechanism, and a distribution method, we use RabbitMQ [30], an imple-476 mentation of the Advanced Message Queuing Protocol (AMQP) [1]. RabbitMQ is 477 a messaging broker, which allows for a common yet generic method for passing 478 messages between components by creating queues for messages, producers that put 479 messages into queues, and consumers that take messages out of the queues.

Figure 12.5 shows the sequence of how messages travel through CUFF using 481 RabbitMQ. The process is initiated at step 1, when the user submits a batch of 482 analysis requests to be done for a case. In step 2, each request is processed by the 483 web server and is reformed to the format understood by RabbitMQ, such that the 484 list of files for a single analysis request is stored in the body of the message, as are 485 the criticality of the request and the tool or algorithm to be used on the files. If the 486 tool or algorithm needs specific command line parameters, these are also stored in 487 the body of the message. After the message has been properly crafted, it is passed 488

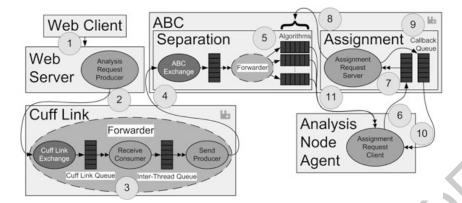


Fig. 12.5 The path of job request messages being passed through CUFF. Single-headed arrows indicate a message being sent to the broker, and double-headed arrows indicate a message being retrieved from the broker. Components with the RabbitMQ logo in the top right corner are running the messaging server

to the Analysis Request Producer. From this point on the message is in the care of 489 the broker until it is delivered to an Analysis Node.

In step 3, the message is sent to the Cuff Link that is running a forwarder utility. 491 In a deployment of CUFF that does not interact with other deployments, this utility 492 is fairly insignificant because it simply passes messages on to the Analysis Block 493 Controller. However, for deployments that interact with each other and allow jobs 494 from one organization to be analyzed on another deployment, this is the mechanism 495 that would be responsible for directing messages to the other deployment's Cuff 496 Link. The forwarding of messages in this manner would depend on the name server 497 of the first deployment knowing how to resolve the domain names of any connected 498 deployments.

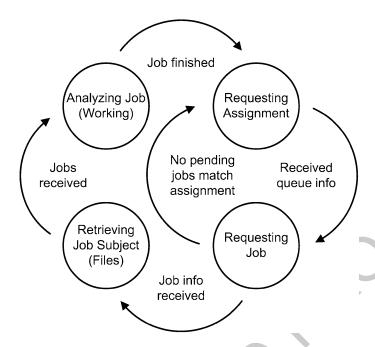
After a message has been received by the appropriate Cuff Link, it is forwarded 500 to the Analysis Block Controller in step 4. In step 5 each message is put into a 501 queue with messages that have both an equivalent criticality level as well as the 502 same analysis type. Messages remain in these queues until retrieved by an Analysis 503 Node Agent.

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Before we proceed, it is appropriate that we discuss the behavior of Analysis 505 Node Agents. As illustrated in Fig. 12.6, each Analysis Node follows a specific set 506 of state transitions which are governed by the agent running on the node. During 507 most of a node's time running in the cloud, it will be working on an analysis task 508 for a job that it has been assigned. When the job is done, the agent sends the results 509 to the Storage component and submits a job assignment request. An assignment 510 indicates to the agent from which queue to take a job. Upon requesting a job from 511 a particular queue, the agent will either be informed the queue is empty, in which 512 case a subsequent assignment request would be made, or it will have the necessary 513 information to retrieve the files from the Storage component and begin the analysis. 514

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**Fig. 12.6** The state transitions for Analysis Node Agents

In step 6, the Analysis Request Client segment of the node agent sends a job 515 request message to the Analysis Request Server on the ABC. During a course of 516 step 7, all assignment request messages are stored in a queue and are processed in 517 the order they were received. The Analysis Request Server then utilizes a read-only 518 interface to the job queues to evaluate the best candidate for the node in step 8.

Next, in step 9, the Analysis Request Server puts this information into the 520 callback queue that was specified as part of the message body from step 6. It is 521 noted that although it technically fulfills the purpose of a queue, we have designed 522 our use of the callback queue to store at most one element. The reason for this is 523 that job description messages sent to it will only be in response to an assignment 524 request message, which will only be sent to the ABC when the node has completed 525 a previous analysis job, at which point it takes step 10 and consumes the contents of 526 the callback queue. Hence, the callback queue is only used because it is required by 527 the broker.

Finally, having received an assignment, the node agent takes one of the jobs 529 from a queue in step 11. At this point, the agent can get the files from the Storage 530 component and begin the analysis.

In addition, for handling the assignment request messages, each message con- 532 tains the necessary information for the ABC to make an appropriate job assignment, 533 namely, the types of analyses the node can perform as well as the desired level 534 of criticality. This turns out to be quite a critical element in this scheme, because 535 it is the node agent that requests what criticality level the job should have that is 536 assigned to it. This means that the anti-starvation requirement is satisfied in the 537 implementation of the node agent, which keeps track of the quantities of jobs 538



completed for each distinct criticality level. Then, before the agent submits an 539 assignment request, it first compares the ratios of completed jobs for each criticality 540 level with the ratio specified by the system administrator and selects the most 541 outlying level to include in the request.

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## **Evaluation of Scheduling Utilities**

The overall viability of CUFF as a digital forensics analysis framework depends 544 on its ability to distribute the work of analyzing evidence by scheduling nodes 545 to be responsible for smaller atomic portions of the analysis work. Therefore, to 546 demonstrate that our method of scheduling jobs is also viable, and to support our 547 claim that the Analysis Block Controller functions efficiently and is scalable enough 548 to facilitate collaboration between examiners, we present our set of tests which 549 simulate real analysis jobs being assigned to nodes and executed.

In terms of execution profile, the procedure of accepting and separating job 551 request messages (steps 1–5 in Fig. 12.5) is remarkably different from the procedure 552 of assigning jobs to nodes (steps 6–11 in Fig. 12.5). The former will occur in bursts 553 of batches as practitioners submit groups of jobs to the system and will not have a 554 continual inflow. The latter will be a steady disbursement of jobs one at a time to 555 nodes as they become ready. Because of such a difference, it is less important that 556 the separation procedure be time-efficient and more important that it provide reliable 557 delivery of every single message to the Analysis Block, whereas the assignment 558 procedure should be likewise reliable but also expeditious to manage queued jobs 559 and respond to each assignment request from nodes so as to minimize their wait 560 time between analyses for the greatest possible throughput.

One challenge in evaluating the performance of the scheduling utilities is the 562 fact that processes are running on completely separate VM instances in the cloud 563 and hence have separate system clocks. For the separation procedure, this proved 564 to make it prohibitively difficult to produce reliable travel times since the path of a 565 single message is linear and does not return to its origin. Even with a Network Time 566 Protocol (NTP) service running on the Cuff Link to try to keep all the components' times synchronized, impossible (i.e. negative) travel times continued to plague our 568 results which are on the order of thousandths of seconds.

The assignment procedure is different, however, because the Analysis Node initi- 570 ates the request and is the final destination, allowing for very reliable measurements 571 that are obtained from the same system clock. Because of this, we will only present 572 the empirical results of the assignment process.

To test how well the assignment process facilitates collaboration, we will equate 574 certain actions of practitioners to the process of automatically retrieving, analyzing, 575 and sharing the analysis results of a task as they are carried out by an Analysis 576 Node. We affirm that this comparison is acceptable for the reason that, whenever 577 the occasion calls for it, practitioners will manually perform these same operations 578 by taking control of an Analysis Node. One difference between the two scenarios is 579 that nodes only retrieve and work on one analysis task at a time. To accommodate 580

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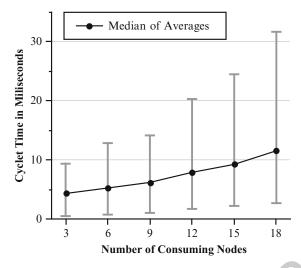


Fig. 12.7 The minimum, maximum, and median values for sets of averages where each set contains average execution times for a number of concurrently running nodes

**Table 12.1** Precise numbers for the evaluations illustrated in Fig. 12.7. Numbers are given in milliseconds

Consumers in group	Minimum average	Median average	Maximum average
3	3.8295	4.31941	5.0408
6	4.5067	5.24674	7.6002
9	5.1649	6.19420	7.9581
12	6.1907	7.89824	12.4222
15	7.0317	9.24481	15.2574
18	8.8711	11.55450	20.1397

this difference, we consider the behavior of groups of nodes that may be coupled 581 with an organization or team of practitioners. In this way, the nodes are not limited 582 in their capacity to collaborate since they can logically pool their resources and 583 results.

To help make the nodes' behavior more realistic, each of the job analysis 585 messages has been given between 5 and 15 fake file descriptors. When an Analysis 586 Node retrieves a job, it simulates the load of processing each file by sleeping for 587 5 ms per file before continuing with its normal execution.

In our test, we created groups of Analysis Node consumers in multiples of 589 three. Timing mechanisms were implemented to measure the completion of steps 590 6-11 from Fig. 12.5, which we call a "request cycle." Each consumer made 2,000 591 synchronous requests<sup>2</sup> with an average cycle time calculated for each set of 100 592 requests. Minimum, maximum, and median values of all nodes in the group were 593 then calculated as illustrated in Fig. 12.7 and as detailed in Table 12.1.

<sup>&</sup>lt;sup>2</sup>Here we mean that each node made synchronous calls while all the nodes ran asynchronously.



We recognize that the evaluations we have presented here focus on a single 595 component in a complex system comprised of many other elements that could have 596 a significant effect on the scalability of our framework. However, with many of 597 these other elements of CUFF still in development, we chose to demonstrate that the 598 most fundamental of all the components is an appropriate method that will facilitate 599 other components' with a high level of reliability and reuse. In doing so, we believe 600 CUFF's abilities can be further extended to take on increasingly realistic analysis 601 tasks.

User Interface 603

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The main purpose of the user interface is to provide a means for the users to 604 take advantage of all of CUFF's features. These features fall into three main 605 categories: evidence browsing and communication, analysis management, and 606 storage management.

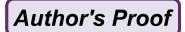
Evidence browsing (extraction) is the task that takes more of an examiner's time 608 during the forensic process than any other task. During this stage of work, the 609 examiner looks through the file system of the acquired evidence, identifies files to 610 be analyzed, studies the results of analyses, and makes decisions based on those 611 results. This is the phase when collaborating with colleagues and subject matter 612 experts is the most beneficial, so it is logical to combine the browsing tools with the 613 collaboration tools into one interface.

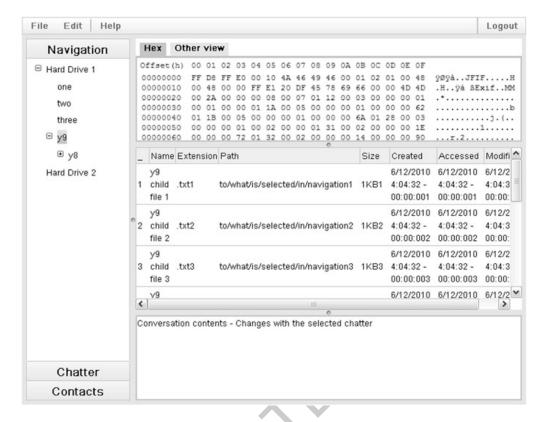
We have created a simple web interface with the Google Web Toolkit that 615 demonstrates one way in which these tools can be combined. As depicted in 616 Fig. 12.8, the left pane provides means for navigating evidence, the user's contacts, 617 and the communication files connected with the case that is currently open. The 618 evidence navigation section is populated by deriving the original file system 619 structure from the DFXML file of the evidence image.

The right side of the interface shows the contents of the selected file, a detailed 621 listing of the currently selected directory, and a space for adding comments about 622 the evidence. At this time, this approach of adding comments is the primary means 623 of collaboration that we have implemented into our system. These comments are 624 stored with the case metadata and analysis results.

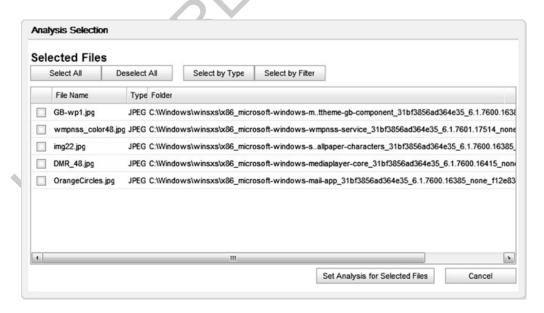
We initially considered implementing a more sophisticated, near-real-time communication mechanism by adopting the Google Wave Operational Transformation 627 algorithm [35]. However, due to the complexity of implementing this algorithm 628 outside its intended use for a deployment of "Wave in a Box" [37] and because 629 of its recent transitory state to become an Apache incubation project [4], that 630 implementation effort still needs further investigation.

During the extraction phase, examiners also need a way to specify what forensic 632 analysis needs to be performed on which files. To do this, the examiner needs access 633 to something that presents the available analysis tools and algorithms for the files 634 that have been selected. Figure 12.9 is gives an idea of what such an interface would 635





**Fig. 12.8** This simple web interface allows users to browse the contents of evidence and communicate with each other



**Fig. 12.9** Using this simple tool in CUFF, users will be able to specify evidence to be analyzed and check on its progress



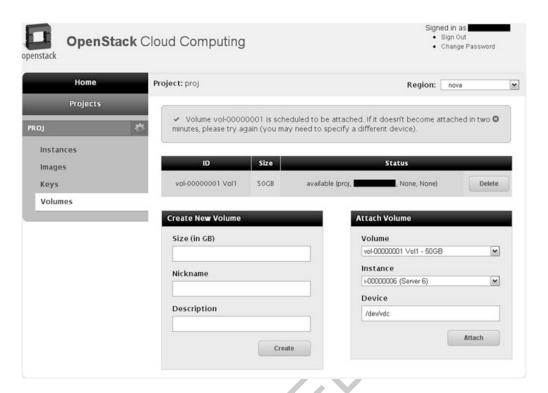


Fig. 12.10 A screen shot of the OpenStack dashboard immediately after connecting a volume to a running instance as /dev/vdc

look like. In the dialog presented to the examiner, the files that have been previously 636 been marked to be analyzed in the evidence browsing window are listed. From this 637 list the examiner can manually select files, select all the files, deselect all the files, 638 select by file type, or select by a regular expression filter. Once the desired files have 639 been selected, the examiner can then click on "Set Analysis for Selected Files" to 640 choose from a list of available analysis algorithms and tools, after which the analysis 641 jobs will be queued into the system.

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Because storage volumes are needed for new evidence images and evidence 643 analysis results, there must be some way for the user to execute the operations of 644 creating volumes and attaching them to instances, at least until such processes can 645 be fully automated by the system. One simple solution to performing these tasks is 646 to utilize the web dashboard provided by OpenStack, which is shown in Fig. 12.10. 647 While using this dashboard for all volume operations will be a bit tedious for jobs 648 of any substantial size, it does provide the needed functionality.

Conclusion 650

In this paper we have discussed the trends of computer crime and the tools to 651 combat those crimes. From these trends we have determined that collaboration 652 among examiners through a secure and robust system would give them a significant 653

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advantage to successfully identify both inculpatory and exculpatory evidence in a 654 timely manner. We set forth our requirements for such a system in a framework 655 based on principles of scalability and interoperability. We then provided details 656 for an implementation of the framework and the additional components that are 657 necessary for the basic operations of a live deployment of CUFF.

For this extended work, we have implemented CUFF on the OpenStack cloud 659 architecture, which provides many needed functions for the system. We also 660 described in detail how the Cuff Link mediates communication between com- 661 ponents and how the Storage component leverages the strengths of OpenStack's 662 storage features. We also presented a potential use of the DFXML data representation format and introduced our XML schema for DFXML to enhance reliability 664 of data types within a DFXML file. In addition, we proposed our approach 665 to scheduling the use of the system's resources through an efficient messaging 666 protocol.

As we continue to improve upon our implementation of all the components in 668 CUFF, we will perform evaluations and usability testing on our system. As part 669 of this effort, we are currently in correspondence with law enforcement agents in 670 multiple locations to ensure that our research is in alignment with the needs and 671 specifications of those for whom these tools are intended.

Another aspect we will consider as we continue our work on this framework is 673 issues dealing with multi-cloud scenarios. We will be exploring means of securely 674 connecting multiple deployments together so as to allow for sharing of resources and 675 analysis tools to a much higher level without compromising the system's compliance 676 with the rules of evidence.

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