

# File Valuation in Information Lifecycle Management

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## ABSTRACT

*Information Lifecycle Management (ILM) stores files according to their value. Therefore file valuation is a very important task in the ILM environment. In this paper we look at how the value of a file can be measured. Instead of traditional methods leading to a classical decimal-value, the method presented leads to a valuation in terms of a "probability of further use". Feasibility of the new method is verified using an ILM simulator.*

## 1. INTRODUCTION

ILM is based on the idea that in an enterprise different information have different values. Valuable information is stored on systems with a high quality of service (QoS). The value changes over time and therefore migration of information is required to cheaper storage systems with a lower QoS. Automated migration makes ILM dynamic. Such automation requires storage systems to understand what files are important at what time so that right policies can be applied. In this point ILM nowadays lacks information valuation methods and tools.

The question is "How is the value of a file measured?". Storage Network Industry Association (SNIA) proposes measuring the value as an amount of money [1].

Other methods express the value as a decimal-value [2]. In section 3 we show how this type of value can be derived from a set of metadata. This method of valuation depends on different factors and has to be defined accurately. Obtaining metadata is not always easy or even possible. Therefore in section 4 we abandon metadata and show how the value can be derived using a probabilistic method. Here the value of a file is calculated from usage information and expressed as a probability of further use. This is a new method which allows valuation depending on the future importance of a file. Section 5 proves the capabilities of the new method using an ILM-simulator. Section 6 applies and combines both methods. The paper ends with a summary and an outlook on our future work.

The essence of this paper is as follows:

1. We present a new method of file valuation
2. We show that this probabilistic method works for ILM systems
3. We combine the new with an "old" method to optimize the performance.

## 2. RELATED WORK

Usage information is used for valuation in other system domains as well. Google uses PageRank algorithm to rank the importance of a web page [3, 4]. A page is ranked based mainly on how many other pages are linked to it. Such links represent a form of usage. They indicate how many other pages are using that particular page. Caching algorithms often rely on data usage information to determine what data are important and hence what to cache in buffers in file systems, databases, and storage controllers [5, 6, 7]. These algorithms cannot be directly applied to our problem due to different design purposes and different target data.

Usage was the focus of Strange, too, who examined the long-term access behaviour on files in an UNIX system [8]. His aim was to identify regularities and patterns which can be applied to automated migration strategies for Hierarchical Storage Management (HSM).

Schmitz has also analyzed the access behaviour of files on a supercomputer to be able to derive an optimal migration strategy [9].

Miller and Gibson examined the access behaviour in further studies in UNIX environments and designed a "file aging algorithm" as a migration rule [10].

The self-\* storage system at Carnegie Mellon University aims to automate storage management tasks through self-managing techniques [11]. It describes how one can classify files based on the automatic learning of file properties using decision tree algorithms.

Chen focused on the file valuation for ILM. He erects value classes which are characterized by a unique set of attributes [2]. The valuation leads to a decimal value which can be normalized to the interval [0;1].

In contrast to other work the valuation presented in this paper offers a percent value. It demonstrates how great the probability of future accesses on a specific file is. Based on the percent value an ILM system can migrate files when their access probability falls below a predefined threshold.

## 3. FILE VALUATION USING METADATA

Metadata are data that describe other data. Therefore, in general, for accurate valuation the more metadata is useable the better the valuation will be. Relevant, but not limiting, factors for ILM are [12]:

- Legislation
- Cost
- User
- File size
- File type

This list could be extended to factors representing the value of knowledge and intellectual properties [13, 14]. These business focused valuation methods require intense human interaction and organizational support. Hence they are often hard to implement.

The five factors mentioned above have different characteristics. For example, some are steady others are discrete, some are string variables others are real variables.

To obtain the information a form has to be filled in. This needs human interaction and makes it difficult and expensive to receive the metadata. Nonetheless we discuss the parameters represented as a mapping which can be implemented into an online form. We will now consider each factor.

*Legislation:* Each file in an enterprise has its own file retention period. In Germany the period can vary between 0 and 10 years. In the American healthcare environment, for example, the period can be up to 100 years [15].

Let  $L(F)$  be the file retention period of file  $F$  determined by legislation.  $L(F)$  is a discrete function:

$$L(F) : F \mapsto L(F) \in \{0, 1, 2, 5, 10\} \subset N_0$$

*Cost:* Each file is important for the enterprise. Its importance is related to the cost originated from the absence of this specific file. Business importance is expressed in a currency (e.g. Dollar or Euro). Either the real value or a relative value is used.

Real values are difficult to obtain. Here the relative value is used. The value might vary between 0 and 10.000.

Let  $C(F)$  be the cost originating from the absence of file  $F$ .  $C(F)$  is a steady function:

$$C(F) : F \mapsto C(F) \in [0;10.000] \subset R_0^+$$

*User:* Each file is intended to be used by specific users within the enterprise. If this factor is to be used for valuation, the users' importance is distinguished between "low", "medium" and "high".

Let  $U(F)$  be the intended user group of file  $F$ .  $U(F)$  is a discrete function:

$$U(F) : F \mapsto U(F) \in \{low, medium, high\}$$

*File size:* The size of a specific file can be used as a factor for valuation, too. The intention is to reduce the needed space on the expensive storage hierarchies. Therefore there is a special focus on large files which represent a great capacity-saving potential. The file size is finite and varies between the values "small", "medium", "big" and "very big". Depending on the enterprise the thresholds are set, e.g. between "big" and "very big" it can lie at 1MegaByte or much higher [9].

Let  $S(F)$  be the size of file  $F$  where  $S(F)$  is a discrete function:

$$S(F) : F \mapsto S(F) \in \{small, medium, big, very big\}$$

*File type:* The file type is determined by the application. In the office environment the most common file types are, for example, "doc", "xls" and "ppt".

A case study conducted in 2004 at an enterprise database identified 21 different file types with the following composition [16].

Let  $T(F)$  be the file type of file  $F$  where  $T(F)$  is a discrete function:

$$T(F) : F \mapsto T(F) \in \{doc, xls, ppt, pdf, rest\}$$

Depending on the enterprise other file types like e.g. "jpg" or "mpg" can be assigned, too.

The five factors are summarized to a vector to derive the value of a file. Let  $V(F)$  be the value of file  $F$  where  $V(F)$  is an n-dimensional function (here  $n=5$ ). It takes into consideration the factors "legislation", "cost", "user", "size" and "type" and

determines the value as a decimal figure. The value is used to assign the files to the different hierarchies in an ILM environment.

$$V(F) : F \mapsto V(F) = \\ V(L(F) C(F) U(F) S(F) T(F)) \in R_0^+$$

Currently  $V(F)$  is only a theoretical mapping. The definite procedure to derive the value from the vector consists of transformations and normalizations. First the string variables ( $U(F)$ ,  $S(F)$  and  $T(F)$ ) are transformed to real variables. Then the real variables are normalized to  $[0;1]$ .

At the end the n-dimensional vector is reduced to a one-dimensional figure. This might happen by simple actions like:

$$V^* := \max \{L^*, C^*, U^*, S^*, T^*\} \in [0,1] \text{ or}$$

$$V^* := \text{mean} \{L^*, C^*, U^*, S^*, T^*\} \in [0,1]$$

This shows that the valuation using metadata works. The advantage of this type of valuation is that no history information is needed. On the other hand, this procedure is neither easy nor cheap.

#### 4. FILE VALUATION WITHOUT METADATA

Instead of metadata we now use usage information for the valuation process. This procedure results from one of the most intuitive metrics for file valuation "if a file is not used for a long time, it is not valuable".

In a case study on a database we provided following results [16]: There were more than 150,000 files on the system and 89 percent of them were not accessed 90 days after creation (see figure 2).

The intuitive method for valuation would lead to the following policy:

*"A file is valuable if it has been accessed during the last 90 days and not valuable otherwise."*

This is a simple way of measurement and one often used in HSM (Hierarchical Storage Management) solutions. It demonstrates that a simple history-based valuation without metadata has a wide-spread acceptance.

Nonetheless this method does not fit to ILM because it only focuses on time limits and not whether the file is needed in the business process which is the intention of ILM [1].

Figure 1. Distribution of file types

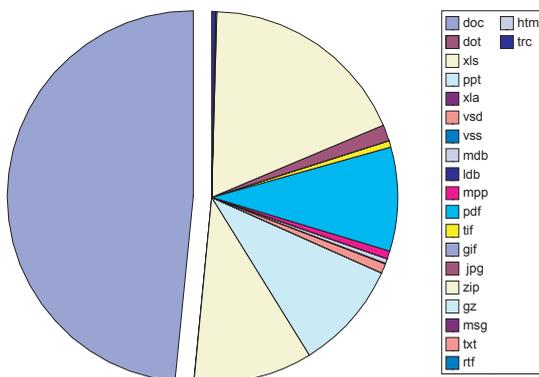


Figure 2. Access probability

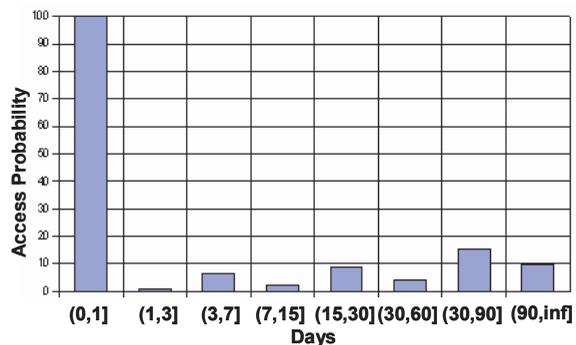


Table 1. Applied distribution functions

Number of accesses	1-6	7-14	15-∞
File type			
doc	W(0,35;3,5)	G(0,32;183)	W(0,35;3,5)
xls	W(0,25;1,1)	W(0,25;1,1)	W(0,25;1,1)
ppt	W(0,38;14,3)	W(0,38;14,3)	W(0,38;14,3)
pdf	W(0,35;3,5)	G(0,32;183)	W(0,35;3,5)
other	W(0,46;27,7)	G(0,29;181)	W(0,46;27,7)

#### 4.1 Case Study

We conducted another case study with the intention of obtaining a predictable measurement of whether or the file is needed in the future.

Our aim was to derive mathematical distribution functions for file accesses for several file types.

We took a sample of 1,000 files from a company's database.

Each file was characterized by the number of accesses per file, the size of the files, the size of the accesses, the age of the files, as well as the file types and access methods.

The file types doc, xls, ppt, pdf and zip are contained most frequently in the sample. The file types avi, cfg, csv, cti, dot, exe, gif, htm, jpg, log, mdb, mmap, mmp, mp3, mpg, mpp, pps, pst, rtf, sql, tif, trc, txt, vsd, vss, wav, wbk, wf2 and xml fall into the category "other".

Most accesses to files in the sample, i.e. 46.22 % of 7,911, are of the "version fetched" type. The access types "View" and "Version added" are represented with 19.20 % and 17.60 % at second and third place. Other frequently occurring access types are "Move", "Reserve", "Unreserve" and "Permission changed". The noticeably more seldom access types under "Miscellaneous" are "Attributes Changed", "Rename", "Copy", "Version Deleted", "Alias Created" and "Generation Created".

We derived distribution functions for file accesses in conjunction with the file type and access history [17]. Either the Weibull-distribution ( $W(\alpha;\beta)$ ) or Gamma-distribution ( $G(\alpha;\beta)$ ) were derived as distribution functions (see table 1).

Using this approach we were able to calculate the future access probability of a file.

The valuation is executed on a percentage basis: "A file is valuable if its access probability is higher than, e.g., 5% and not valuable otherwise".

This is a new quality of valuation using only the file type and the access history. This information is provided by the storage systems and does not need further metadata or user interaction.

#### 5. PROOF OF CONCEPT

We implemented a simulator to analyse that the valuation based on "percentage of further accesses" can be used for ILM. Figure 3 shows an example using 5 hierarchies observed over 4,000 days. The lifecycle is illustrated graphically for one single file.

In figure 3 you can see that between day 1,300 and day 1,500 after generation of the file the access probability sinks remarkably. Therefore the file is moved to the lowest storage hierarchy in several steps. After approximately 1,800 days the file is accessed so that it must be shifted to the highest storage level again.

After about 3,800 days the information is again at the lowest level. The lifecycle follows the typical course of a file with a periodical access sample.

A graphic evaluation does not make sense for the simultaneous analysis of several files. Therefore examination of the relative capacity-need per hierarchy has to be performed.

Figure 4 depicts how an ILM system with 3 hierarchies and the described valuation works.

Figure 3. Migration of a file according to its value measured in "% of further accesses"

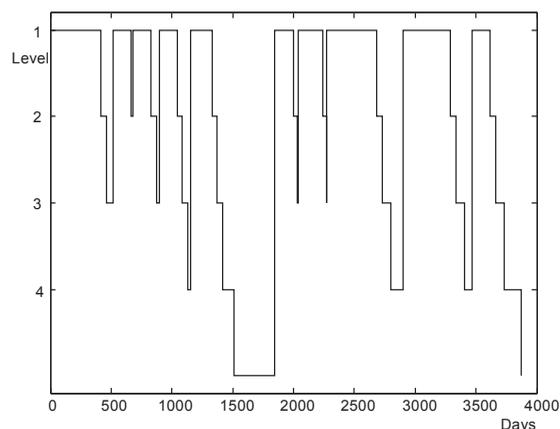
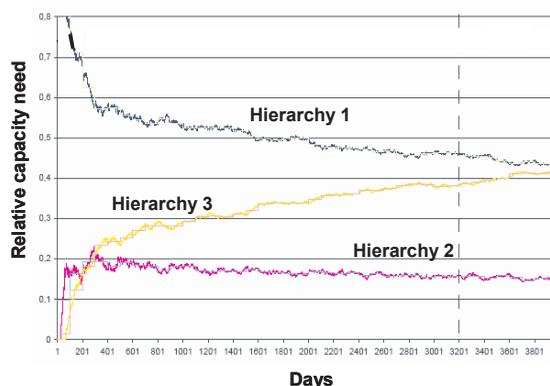


Figure 4. Relative capacity-needs in a 3-dim. ILM system



Since the method looks at the access history, it needs some time to stabilize the capacity-need in an ILM system.

#### 6. COMBINATION OF THE DERIVED VALUATION METHODS

When metadata is available, it is advisable to use it in combination with the probabilistic method. Since the effort in dealing with metadata is quite high, it is advisable to use metadata only for the initiation of an ILM scenario.

There are two options for initiating an ILM scenario (see figure 5):

- Option 1: Store all files on the highest hierarchy at the beginning.
- Option 2: Valuate all files and presort them into the related hierarchy at the beginning.

Option 1 does not require metadata, of course. The files are stored in the highest hierarchy. Their valuation is done on the basis of file access patterns during the ILM process as shown in figure 4.

The advantage of option 2 compared to option 1 is that the capacity need for hierarchy 1 is below 100% from the beginning. This means that money is saved earlier on.

Figure 6 shows a simulation run in which the files were presorted according to option 2.

We notice that presorting has positive effects on reaching system stability.

Figure 5. Starting options in an ILM system

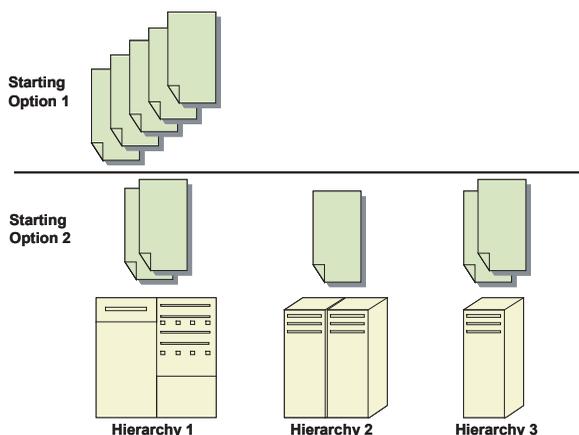
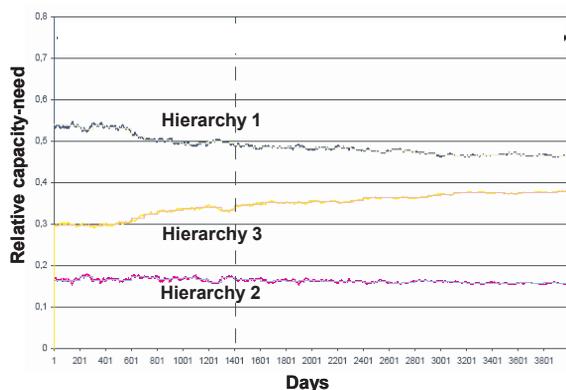


Figure 6: Relative capacity-needs in a presorted 3-dim. ILM system



**7. SUMMARY AND OUTLOOK**

Proper information valuation is the first step towards ILM automation. Existing valuation methods either use metadata or look at the history and generate a value in terms of “amount of dollars” or “a decimal figure within an interval”.

We demonstrated that the future access of a file can be predicted from observed access data and that this can be used as a metric for file valuation.

The value of a file is its percentage of further accesses. This is a new way of valuation. The advantages are that it is simple, does not need metadata and fits to ILM automation. Furthermore it is more sophisticated than the simple HSM-time-limit approach.

Application of the new method was shown by means of an ILM simulator. Therefore the first step towards automation is done.

Since the method looks at the access history, it needs some time to stabilize the capacity-needs in an ILM system.

This period can be shortened by presorting the files over the hierarchies. We showed that in combination with an initial valuation according traditional methods using metadata earlier cost gains in ILM-systems can be achieved.

In our future research we will continue to look at ILM automation. We will focus on the aspect of policy definition. In order to compare different policies, the existing simulator will be optimised.

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**ENDNOTE**

<sup>1</sup> \* indicates that transformation and normalization have been executed

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