

## Adaptive Content Presentation for the Web

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**Abstract.** In this chapter we describe techniques for adaptive presentation of content on the Web. We first describe techniques to select and structure the content deemed to be most relevant for the current user in the current interaction context. We then illustrate approaches that deal with the problem of how to adaptively deliver this content.

### 13.1 Introduction

Previous chapters in this book have described types of adaptation for Web-based systems that include adaptive navigation support (see Chapter 8 of this book [8]), adaptive search (see Chapter 6 of this book [39]) and personalized recommendation of items of interest (see Chapters 9 [47], 10 [42], 11 [49], and 12 [9] of this book). In this chapter, we will focus on an additional type of adaptation widely known as *adaptive presentation of content*: how to present Web-based content in a manner that best suits individual users' needs. This type of adaptation involves determining, based on the user and context, what information the system should present and how the information should be organized and displayed. While adaptive presentation of content can serve many purposes, as we will demonstrate throughout the chapter, it can also complement several of the adaptation types discussed in previous chapters. For instance, the content of Web pages pointed to by a tailored link in a system that provides adaptive navigation support (Chapter 8 of this book [8]), or returned by adaptive search (Chapter 6 of this book [39]), can be modified to highlight the parts that are more interesting for the current user. Similarly, the description of the items returned by a recommender system (see Chapters 9 [47], 10 [42], 11 [49], and 12 [9] of this book) and can be adapted to play up the items' features that are more relevant to the user's needs, or changed to be more suitable to the user's level of familiarity with the items.

The focus of this chapter will be on computational techniques necessary to provide the user with a tailored presentation of content, rather than implementation details and technologies. Also, the chapter is not limited to techniques currently used in adaptive Web-based applications. It aims to suggest areas of future research by discussing alternative approaches that have a strong potential to augment the set of existing techniques for adaptive presentation on the Web.

The process of adapting content to specific user needs comprises two sub processes: content *adaptation* and *presentation*. Content adaptation involves deciding what content is most relevant to the current user and how to structure this content in a coherent way, *before* presenting it to the user. The second sub process of *content presentation* involves deciding how to most effectively adapt the presentation of the selected content to the user.

The chapter is structured as follows. In section 13.2, we address techniques for content adaptation. Although traditionally these techniques required the existence of pre-crafted versions of the relevant content, new techniques are emerging which can automatically adapt content from abstract knowledge sources. Given that the latter lead to greater flexibility and robustness, our discussion focuses on these. In section 13.3, we discuss techniques for content presentation. We first introduce techniques that deal with the problem of how to present this content so that user focus/attention is drawn to the most relevant information (possibly defined by using any of the techniques described in section 13.2) while still preserving the contextual information that can often be provided by content of secondary importance. We then discuss techniques to decide which media/modality to use to best convey the selected content.

## 13.2 Techniques for Content Adaptation

Content adaptation involves identifying the content most relevant to a given user and context (jointly referred to as the interaction context), as well as how this content should be organized. Relevant properties of the interaction context can include the user's preferences, interests, and expertise, as well as the presentation goals. Content adaptation of Web pages can be characterized along the following key dimension: the nature of the content provided as input. Along this dimension, we first briefly describe two rather simple approaches in which adaptation is achieved by selecting appropriate canned pages or page fragments. These approaches are referred to in the literature as *page* and *fragment variants* respectively, and they have been extensively discussed in previous surveys (e.g., [32]). After a brief description of page and fragment variants, we provide an in-depth discussion of more sophisticated approaches to content adaptation in which the input is abstract information, since to the best of our knowledge, these approaches have never been covered in detail in any previous survey on adaptive hypermedia.

### 13.2.1 Approaches Based on Page and Fragment Variants

The simplest form of content adaptation is the page-variant approach [32]. Here, the input of the adaptation process consists of different versions of each page that is to be adapted along with a model of the interaction context. These versions have to be written in advance. At runtime, the adaptation mechanism selects and presents the page version that is most appropriate to the current interaction context. Clearly, this approach does not scale up to complex adaptation. If several aspects of the page must be adapted in many different ways, an unmanageably large number of variants need to be written. Nevertheless, in some domains, where only high-level adaptation is needed, this approach has been effectively applied. For instance, in the ORIMUHS system

[14] page variants are applied to support user interaction in two complex software systems: a CAD modeler and a medical application. Page variants are also applied in the KBS Hyperbook system [24] to develop educational courseware on Java programming.

Moving up in the ladder of adaptation complexity, we have the fragment-variant approach. In this approach, the adaptation is performed at a finer level of granularity. More specifically, the page presented to the user is not selected from a pool of fixed pages. Rather, it is constructed by selecting and combining an appropriate set of fragments, where each fragment typically corresponds to a self-contained information element, such as a text paragraph or a picture. As with the page-variant approach, these fragments are written in advance. Two common strategies for fragment variants are: *optional fragments* and *altering fragments*. In optional fragments, a page is specified as a set of fragments, where each fragment is associated with a set of applicability conditions. At runtime, the page is generated by selecting only those fragments whose conditions are satisfied in the current interaction context. For instance in [16], different optional fragments are selected depending on the user's knowledge, interests and abilities. Altering fragments are rather different from optional fragments. In altering fragments, a page is specified as a set of constituents, and for each constituent there is a corresponding set of fragments. At runtime, the page is created by selecting for each constituent the fragment that is most appropriate in the current interaction context. Altering fragments are applied, for instance, in the AHA system [13], in which different presentations of the same entity can be selected depending on whether the target user has the necessary background knowledge.

In general, a noticeable disadvantage of fragment variants compared to page variants is that the selection and assembly of a suitable set of fragments may involve a substantial overhead at runtime. Furthermore, it may sometimes be difficult to combine the set of independently selected fragments into a coherent whole. On the other hand, the key advantage of this approach is that, once a set of fragments and their applicability conditions have been written, a large number of pages can be automatically generated to cover a corresponding large number of interaction contexts. For pointers to specific techniques to implement the fragment-variant approach the reader should refer to [32].

Note that because in the two approaches above the units of content adaptation are either whole pages or predefined page components, the two sub processes of content adaptation and presentation actually coincide. That is, the decision of what content is most relevant to the user (i.e. the page to be displayed) uniquely identifies what will be presented to the user. On the one hand, this simplifies and speeds up the complete adaptation process. On the other hand, it reduces flexibility because it eliminates the possibility to further tailor the information through adaptive presentation techniques once the first level of adaptive content presentation, content selection, has been achieved, as we will see in section 13.3.

### 13.2.2 Approaches Based on Abstract Information

Although many adaptive Web systems have been designed in recent years by relying only on page or fragment variants, in this section we describe more sophisticated adaptation techniques that allow a system to reason about the input content and the

interaction context, both of which are expressed in more abstract terms. These techniques permit the adaptation to be more flexible, robust and scalable. Notice that part of the research on sophisticated content adaptation has been developed in the field of Natural Language Generation (NLG) [43], which investigates how natural language text can be generated from abstract non-linguistic information.

Sophisticated content adaptation, also called tailoring in NLG, requires an abstract representation of the domain from which the content is selected, as well as the features of the interaction context to which the content is tailored. Several formalisms have been used in the literature, including:

- *Traditional Knowledge Bases* [46] expressing domain entities and relationships between them. For instance, one application of the ILEX system [40] generates tailored jewel labels by relying on a large object-centered knowledge base about jewelers, materials, designers, etc. This knowledge base includes both abstract propositions, such as the fact that a necklace is a jewel, and specific propositions, such as the fact that a particular jewel was made in Birmingham in 1905.
- *Bayesian Networks* [46] expressing probabilistic relationships between random variables representing the domain. For instance, one application of the NAG system [33],[55] generates arguments about the expected rate of a researcher's future publications by relying on a Bayesian Network. This network specifies probabilistic relationships between the publication rate of a researcher and the factors that influence it, such as the strength of the institution from which the researcher graduated (e.g., the stronger the institution, the higher the likelihood of a high publication rate).
- *Preference Models* [46] expressing the user's preferences about different aspects of the domain. For instance, one application of the GEA system [10] generates user-tailored arguments on whether the user will like/dislike a given house by relying on a model specifying what aspects of a house the user cares most about (e.g., location, amenities). The PRACMA system [29] also employs a model of user preferences to tailor its description of an individual recommended item (e.g., a car) by focusing on the aspect (e.g., price) that will have the largest impact on the user's overall evaluation of that item.

Depending on the application, the same or different formalisms can be used to represent the domain model and the interaction context. For instance, in NAG [55], a system for generating factual arguments (claiming that something is or is not the case), both the domain and user model are represented as Bayesian Networks. Similarly, in HYLITE+ [5], a system for generating adaptive hypertext encyclopedia-style explanations, both the domain and the user models are expressed as traditional knowledge bases, more precisely as conceptual graphs [50]. In contrast, in GEA, a system for generating evaluative arguments (claiming that something is good vs. bad), the domain model is represented as a traditional knowledge base while the user model is expressed as a value tree [46], which is a preference model commonly used in decision theory.

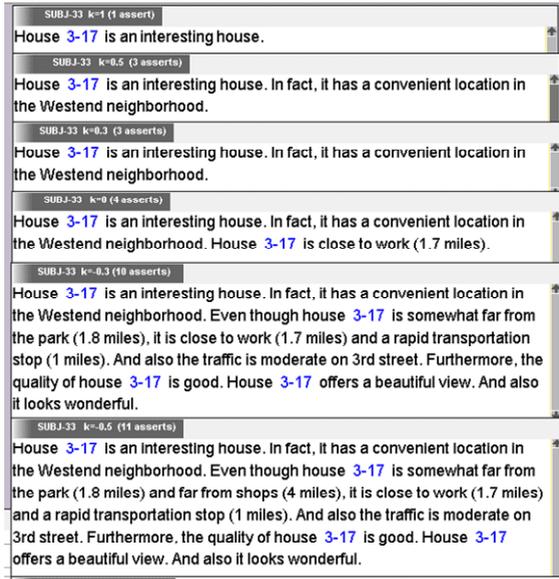
The process of sophisticated content adaptation involves the two conceptually distinct phases of *content selection/determination* and *content structuring*, also jointly referred to as content planning. Although we will describe them separately to simplify the presentation, it should be noted that content selection and structuring are often implemented as one single process that simultaneously performs both phases [43].

**Content Selection.** During content selection, a subset of the domain knowledge is identified as relevant for the current user and situation. Strategies for content selection rely on domain-specific knowledge to different degrees. For instance, the content selection strategy used in STOP, a system for generating smoking cessation letters, is quite domain specific as it refers to psychological knowledge about addictive behavior and smoking [44]. In contrast, the content selection strategy used by the GEA system does not rely on any domain-specific knowledge (as we will see later in this section) and can be therefore applied in any domain [10]. Because of their generality, in this section we focus on strategies that are primarily domain-independent. For a discussion of more domain-specific strategies and in particular of how they can be acquired, the reader should refer to [43].

In practice, most domain-independent strategies for content selection compute a measure of relevance for each content element (i.e., fact) and then use this measure to select an appropriate subset of the available content. Content adaptation is achieved by having this measure of relevance take into account features of the current user and context. For illustration, let's consider three systems that provide a representative overview of how the measure of relevance can be computed and how it can be used for content selection.

*The Intelligent Labeling Explorer (ILEX).* We start with ILEX [40], a system for generating contextually-relevant hypertext descriptions of objects (e.g., museum artifacts, computer components). In ILEX, the measure of relevance for content selection combines a measure of structural relevance of a knowledge element/fact with its intrinsic score. Structural relevance takes into account the structure of the domain knowledge base - a semantic net. More specifically, structural relevance is computed starting from the focal entity (i.e., the entity being described) by considering two basic heuristics: (i) information becomes less relevant the more distant it is from the focal object, in terms of semantic links; (ii) different semantic link types (e.g., GENERALISE) maintain relevance to different degrees. The intrinsic score of a knowledge element combines numerical estimates of three factors: (i) the potential interest of the information to the current user, (ii) the importance of the information to the system's informational goals and (iii) to what degree the user may already know this information. Once the two measures of structural relevance and intrinsic score have been computed, they are combined in a single measure of relevance by straight multiplication.

In ILEX, the content selection strategy is then to return the  $n$  most relevant knowledge elements. However, if the selection process based on relevance cannot find a sufficient number of knowledge elements, additional content selection routines are activated. For instance, one technique applied by ILEX is to identify an entity which is sufficiently similar to the focal entity, so that an interesting comparison between the two can be also selected for presentation. In general, when the goal of a content selection component is to return a fixed amount of content, it may be necessary to supplement the main selection strategy with a set of ancillary strategies.



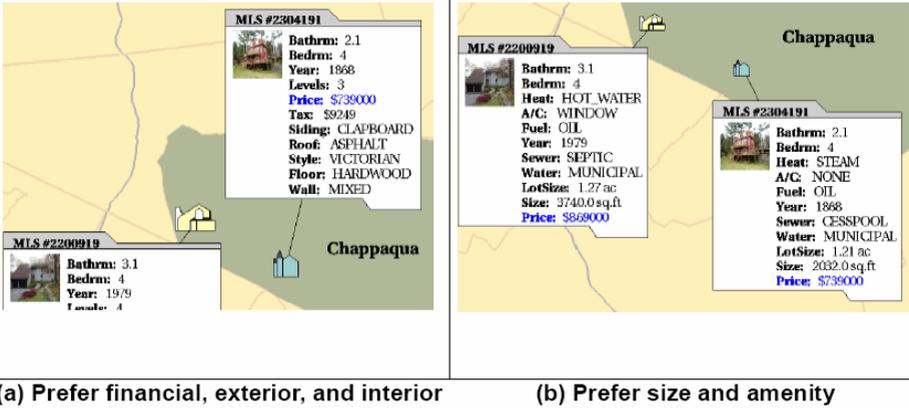
**Fig. 13.1.** GEA system [10]: arguments about the same house tailored to the same user containing an increasing (or equal) amount of content.

*The Generator of Evaluative Arguments (GEA).* As mentioned before, GEA is a system for generating evaluative arguments (claiming that something is good vs. bad). Here the measure of relevance is computed by applying a quantitative model of the user's preferences to the entity being evaluated. Generally speaking, GEA's user model relies on the notion that if something is valued, it is valued for multiple reasons. More specifically, for each user, the model specifies a decomposition of the user's overall assessment of entities in a given class (e.g., houses) into a hierarchy of aspects of the entities (e.g., location, number-of-bedrooms). The model hierarchy is annotated with numerical weights and functions that specify the relative importance/preferability of each attribute and domain-value (e.g., two bedrooms) for the particular user. Once the model is applied to an entity, it is possible to compute for each attribute how much its evaluation contributes to the overall evaluation of that entity for the current user. Based on this, a measure of relevance is defined by assuming that an attribute is relevant either because of its strength or because of its weakness in contributing to the value of the entity. For instance, if distance-from-work is an important attribute for the current user, this attribute will have high relevance with respect to the evaluation of a house that is very close to the user's workplace (because of its strong contribution), as well as to the evaluation of a house that is very far from it (because of its weak contribution).

Once the relevance of all attributes is assessed, the content selection strategy in GEA is to return for each level in the user model hierarchy, only those attributes whose relevance is greater than a (customizable) threshold. By setting this threshold to different values it is possible to generate, in a principled way, arguments that con-

**J3:** Show houses under \$1M in Chappaqua

**RIA:** I found 2 houses under \$1M in Chappaqua



**Fig. 13.2.** RIA system [53]: examples of queries and user-tailored responses for two users with different preferences.

tain different amounts of user-tailored relevant content, as shown in Fig. 13.1 (see [10] for details).

*The Responsive Information Architect (RIA).* In RIA [53], a multimedia conversation system to support information-seeking tasks, content selection is formalized as an optimization problem. The goal is to identify the most desirable subset of data dimensions (e.g., price and style in the real-estate domain) in the current interaction context. The desirability of each data dimension is computed as the linear combination of a large set of feature-based metrics that characterize how important the dimension is with respect to the interaction context. Most of these features are labeled as content-relevance features and include features of the data (e.g., inherent importance of the dimension in the domain, such as the importance of price in the real-estate domain), and features of the user (e.g., relation of the dimension to the user's interests, such as how much the user cares about price). There are also features relating the dimension to the user request (e.g., the user may be requesting information on a particular dimension, such as explicitly requesting information on price) and to the interaction history, which tries to maintain a coherent presentation of dimensions across multiple queries. So, in RIA, desirability is fundamentally what we have been referring to as a measure of content relevance. Once all the available data dimensions have been assigned their desirability, RIA's content-selection strategy returns the set of data dimensions such that their overall desirability is maximized and their cost is within the given space and time allocated for the target presentation. Special techniques are introduced to take care of dependencies between data dimensions. For instance, some groups of dimensions are considered as a single bundle (e.g., number of bedrooms and number of bathrooms) and are either all included or all excluded from the final presentation.

Fig. 13.2 shows how RIA, given the same query, selects different data dimensions for two users with different preferences in the real estate domain.

To summarize, we have seen three ways in which a measure of content relevance can be computed and three prototypical ways in which such measures can be used for content selection. In general, there is no accepted set of guidelines in the field to choose the most appropriate measure of relevance and selection strategy given a target application. For any new application, designers should first consider solutions presented in the literature (of which the systems we have just described provide a representative sample), and devise alternative solutions only if no existing one is satisfactory.

**Content Structuring.** Once the most relevant content elements are selected they must be organized in order to be effectively communicated/presented. This involves not only ordering and grouping them, but also specifying what discourse relations (e.g., contrast, evidence) [31] must hold between the resulting groups. Schemas [43] are the method of choice to accomplish all these tasks and are commonly implemented with task-decomposition planners (technically referred to as HTN planners [46]).

With respect to content selection, we provide a much more limited treatment of content structuring because adaptation of the latter is rather less common than adaptation of the former. One form of structure adaptation is to rely on the measure of relevance used in content selection for ordering the selected content elements. For instance, in GEA the selected elements are ordered according to the measure of relevance by following principles from argumentation theory [10]. Another form of content structuring adaptation is the selection of the discourse relations. For instance, in GEA a given fact can be selected as supporting or contrasting evidence depending on the user's preference for that fact, as determined by the evaluation of the user model.

Note that any structuring information derived from this phase of content adaptation can serve as a guide to decide how to actually present the selected content to the user. The various techniques for adaptive content presentation that we will overview in the next section can be used to adaptively render the information defined by content structuring. For instance, techniques for content emphasis can be used to express relevance information implicit in content ordering. Discourse structure can be used to identify portions of texts that can be made available on demand instead of being displayed up-front [41]. Finally, relationship information can be used to define which medium to use to display the related content (e.g., graphics to highlight quantitative relationships such as a comparison between two sets of numbers [30]). In practice, most previous and current work on adaptive content presentation relies on simpler text-based fragments, which are selected according to a simple user model. However, we believe that work on integrating adaptive content structuring with adaptive content presentation promises to be of great value for the advancement of adaptive content presentation on the Web.

We conclude this section on content adaptation from abstract information with a brief discussion of issues related to knowledge acquisition and evaluation. While all of the techniques for content adaptation from abstract information discussed so far have been engineered by researchers, recent and promising work is exploring how content adaptation strategies can be learned from user feedback on sample (multimedia) presentations [20]. This work relies on machine learning techniques that have been already successfully used in other NLG adaptation tasks [51]. As for evaluation,

recent years have witnessed a surge of interest in empirically testing techniques for content adaptation from abstract information. Human judges [11], human designers [53] and task efficacy (e.g., [5], [10]) are the three basic methods that have been applied to evaluate such techniques. These methods are described under the category "controlled experiments" in Chapter 24 of this book [18], which provides a comprehensive overview of all empirical and non-empirical evaluation methods for adaptive Web systems.

## 13.3 Techniques for Content Presentation

In the previous section we discussed techniques to identify and structure the content most relevant to the interaction context. Here, we focus on techniques to effectively present content once its degree of relevance and its structure have been determined by content adaptation. In particular, we present techniques that decide how to present content based on its relevance, and techniques to select the type of media most appropriate to deliver the content, given the interaction context.

### 13.3.1 Relevance-Based Techniques

Most of the techniques that we categorize here as relevance-based were introduced in [7] and [32] as ways to manage canned content fragments. Following [52], in this section we will discuss them along two general dimensions, which we see as critical to both canned and generated content:

- Maintaining *focus*, i.e., how these techniques emphasize the content that has been classified as most relevant for the current user.
- Maintaining *context*, i.e., if and how they allow for access of the less relevant content so as to preserve the contextual information that it may provide.

There is an obvious tradeoff between these two dimensions: context is more easily maintained if much of the original content is visible to the user. However, the more content is shown, the higher the chance of generating information overload and reducing attention to the most relevant information, defeating one of the very reasons for having adaptive hypermedia in the first place.

The techniques we present here can be grouped in two main categories, depending upon how they address the context-focus tradeoff: Priority on Focus and Priority on Context.

**Priority on Focus.** All of the techniques in the Priority on Focus category choose to maximize focus by (a) showing the user only the content that is deemed most relevant, and (b) precluding access to the rest of the content. They include not only the fragment-variant techniques that we discussed in the previous section (i.e., optional fragments and altering fragments), but also any strategy for sophisticated content selection in which only the most relevant knowledge elements are presented to the user (see for instance Fig. 13.1 and Fig. 13.2 in section 13.2).

In addition to potentially losing contextual information when limiting the content the user can see, Priority on Focus techniques suffer from two main drawbacks:

- They are highly impacted by the validity of the adaptation mechanism, as the user has no way to recover from bad adaptation.
- They do not allow for user control, one of the dimensions that defines usability in human-computer interaction and that significantly influences acceptance of adaptive interfaces ([27] and [28]).

Priority on Focus techniques seem to be mostly used in Adaptive Educational Hypermedia (see [13] and [38]), possibly because these systems are less subject to the above drawbacks. The pedagogical nature of the interaction makes it both easier for the system to create an accurate user model of relevant user traits, and possibly more acceptable for the user to have limited control over the computer tutor's adaptation decisions.

**Priority on Context.** The Priority on Context category includes the techniques known as (1) *stretchtext*, (2) *dimming fragments*, (3) *colouring fragments*, (4) *sorting fragments* and (5) *scaling fragments*. The first four techniques have been around for a while and were previously addressed in adaptive hypermedia reviews (e.g., [7]). The scaling fragments technique is a more recent attempt to adapt the well known *fish-eye* visualization technique [17] to content organization in adaptive hypermedia. For this reason, we will describe it in slightly more detail than the other four. In general, all techniques presented in this section try to preserve the context around the most relevant content by providing different ways to make the less relevant information visible without distracting the user from the primary content. They differ, however, along the following dimensions: 1) whether or not the surrounding context is visible, 2) whether or not they permit structural information to be maintained, and 3) whether or not they can convey different levels of relevance and/or priority information. We first describe the techniques and then discuss how they differ along these key dimensions.

*Stretchtext*, arguably the most well established of the techniques, relies on placeholders to signal the presence of and allow access to information of secondary importance (see for instance [6] and [26]). Usually the place holders are short headers summarizing the hidden information, as shown in the screen shot of the Push system [26] in Fig. 13.3 (see items labeled as IE – Information Entities – and Hotlist). While this technique has been mostly used with text (hence the name), researchers have started generalizing it to the adaptive presentation of multimedia content, for instance in Interactive TV applications [36].

*Stretchtext* preserves focus by hiding the less relevant content when a page is first presented. In contrast, the remaining four techniques *deemphasize* rather than hide the less relevant content. *Dimming* deemphasizes the less relevant content by fading its color [6]. *Colouring*, which has traditionally been used for adaptive navigation support (see Chapter 8 of this book [8]), highlights the more relevant content using one or more colours. *Sorting*, also most frequently used for adaptive navigation support, deemphasizes less relevant content through fragment ordering. [25]. *Scaling* deemphasizes by reducing size. In scaling, size increases as a function of the Degree of Interest (DOI) of each content fragment for the user [52], as assessed by a similarity measure between vectors representing the user's focus of interest and the content of each available fragment. In this respect, scaling is a variation of the *fish-eye* visualization technique, with the following differences:

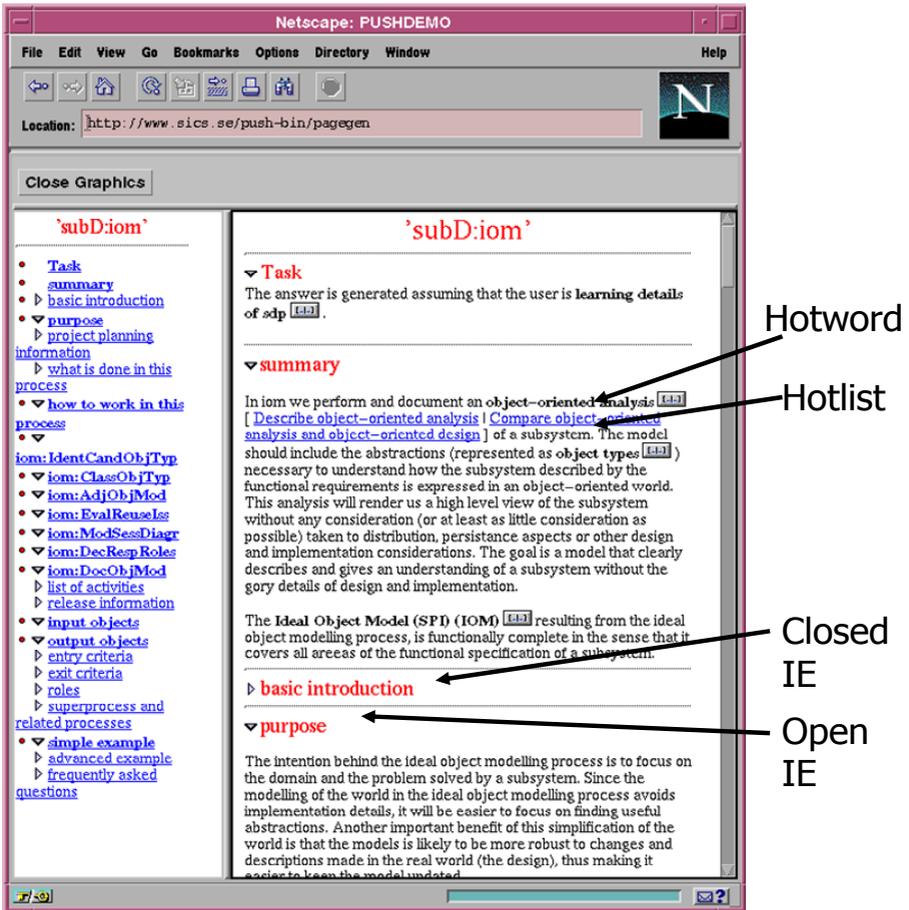


Fig. 13.3. Example of stretchtext-based adaptation in the Push system [26].

- In a traditional fisheye visualization, there is a unique focal point, based on the user's current focus of attention. In scaling, there can be multiple focal points based on the user's focus of interest.
- In traditional fisheye views, DOI (and thus content size) decreases with geometrical distance from the focal point, while in scaling, DOI decreases with *semantic* distance.

Fig. 13.4 shows an example of scaling from [52]. Here the user's focus of interest is assessed to be theater, thus paragraphs with different degree of relation with this topic are presented in different font sizes.

Table 13.1 compares the five techniques discussed in this section. Of the five techniques, stretchtext is the only technique that does not make the surrounding context visible, while sorting is the only one that doesn't maintain structural information. As a result, sorting is suitable only for fragments that are not structurally

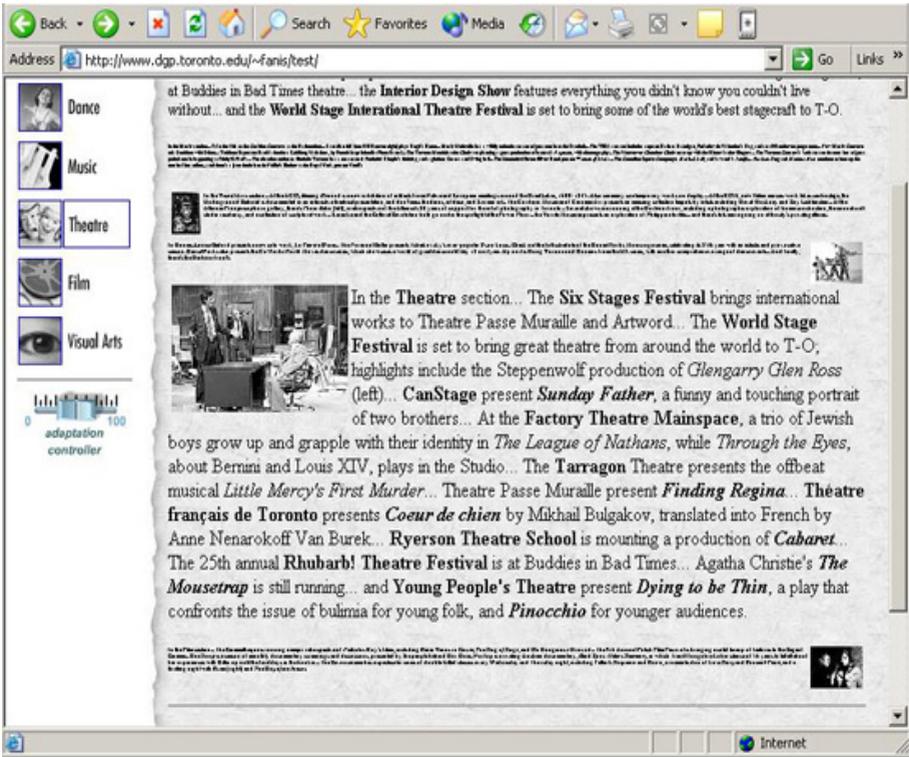


Fig. 13.4. Example of scaling-based adaptation from [52]

Table 13.1. Comparison of Priority on Context techniques

Technique	Context Visible	Structural Information Preserved	Priority Conveyed
Strechtext	No	Yes	No
Dimming	Yes	Yes	No
Colouring	Yes	Yes	To a limited degree
Sorting	Yes	No	Yes
Scaling	Yes	Yes	Yes

related, such as bulleted lists or self-contained sub topics. In terms of conveying priority information, the existing incarnations of dimming equally deemphasize all of the less relevant content; the same is true of stretchtext. In comparison, scaling, sorting and colouring do allow priority information to be displayed. In scaling, size conveys different degrees of relevance. With sorting, the order of appearance on the page conveys the relative relevance. With colouring, more than one colour can be used; however, there is a limit to the number of colours to which the user will be able to attribute meaning.

Although all five techniques in this category do provide more contextual information than those in the Priority on Focus category, their relative effectiveness in different applications is still an open question.

The scaling technique preserves distinctive structural elements of the deemphasized information (such as pictures, layout, number and relative length of paragraphs), and thus has better potential to provide contextual cues than stretchtext. Tsandilas and Shraefel [52] conducted a preliminary study to compare stretchtext and scaling, but the small number of subjects does not allow drawing general conclusions. They found no difference in terms of task completion time, but identified a potential interaction between technique effectiveness and page size. Because stretchtext generally presents less content, it performed better on large pages, where scaling required more scrolling to access relevant content. However, 4 out of the 6 subjects in the study gave a higher overall score to scaling because they felt it provided better information on the content of the deemphasized paragraphs.

Like scaling, dimming and colouring also preserve distinctive structural elements of the available content, but do not reduce page size (like stretchtext and scaling do). Thus, they may require additional scrolling to access the content of interest, potentially reducing focus. Finally, sorting does not preserve structure, however, placing the more relevant material at the top of the page has the advantage that scrolling is necessary for only the less relevant content. Further evaluations are needed to better understand the pros and cons of the above techniques.

Even as more empirical results become available, practitioners interested in adopting these techniques should be aware that their effectiveness depends on a number of design elements that should be carefully tested before drawing general conclusions on each technique's overall effectiveness. These elements include:

1. Quality of the headers used to indicate the presence of stretchtext ([26] and [48])
2. Cost associated with reading the deemphasized content (e.g., fatigue generated by reading small or faded text)
3. Presence and quality of ways to summarize deemphasized content so that a user can get contextual information without reading it (see for instance the mouse-over glosses described in [52])
4. Presence and effectiveness of mechanisms provided to the user to change content emphasis (e.g., double clicking on deemphasized content to bring it in focus [52])

In relation to the first element above, we describe a technique that, although currently not used in adaptive hypermedia, has an interesting potential for enriching the Priority on Context category. This technique, known as *summary thumbnail* [35] has been devised to address the problem of how to display Web pages designed for desktop-sized monitors on small screen devices. One common way to address this problem is to rescale the page to fit the width of the small screen, thus creating what is known as a *thumbnail view*. The idea is that the user should use this view to rapidly identify the content of interest and then use provided zooming mechanisms to view it. The problem is that often text in the thumbnail view becomes unreadable, forcing the user to resort to zooming to browse the content. Summary thumbnail addresses this problem by adding fragments of readable text to the thumbnails (as opposed to shrinking the

text of the original page). The text is automatically generated from the original Web page, by either removing common words (as defined in a standard word frequency list) or by cropping paragraphs, so as to maintain the total number of lines and overall page layout.

Lam and Baudisch [35] showed that this technique generated better user performance and satisfaction in a browsing task, compared to two other standard techniques for page reduction. This indicates that having readable, although incomplete and possibly not fully coherent text is an effective place holder for hidden content. Thus, this technique could be adapted as a variation of stretchtext for adaptive content presentation, where place holders for hidden text are summary thumbnails. However, it should be noted that in Lam and Baudisch's experiment [35], all content was equally reduced. In order to verify the applicability of this technique to adaptive approaches, it will be necessary to test whether it remains effective when summary thumbnails are used in combination with fully-displayed content during adaptive presentation.

### 13.3.2 Techniques for Media Adaptation

The previous sections discussed techniques to adaptively select, structure, and present relevant information. Presentation has been addressed, however, with respect only to the problem of how to highlight relevant information and how to allow the user access to relevant context. Here we will address a different form of presentation tailoring: adapting the medium (e.g., text, graphics, spoken language) through which the selected information is conveyed to the user. We begin by discussing factors that can influence a system's choice of media. We then provide illustrative examples of adaptive hypermedia systems that adapt the medium through which information is presented. Finally, we discuss at a conceptual level two common approaches to media adaptation: the rule-based approach and the optimization approach.

**Factors Relevant for Media Adaptation.** The following is a description of the types of factors that a system may want to consider when deciding how to adapt the media:

- *User-Specific Features:* Relevant user features include preferences, abilities and accessibility issues. Users may have preferences for receiving information in different modalities, for example, a user may explicitly request the information to be presented in a graphical way [54]. In terms of abilities, the user may be better able to reason about information presented using a given medium. For example, if the user has poor language abilities, the material would be better presented visually [21]. In contrast, a visually-impaired user should not be presented with information in this manner, but rather through speech (e.g., [16]).
- *Information Features:* Given the presentation goals, certain types of information are better presented using specific media. For example, graphics should be used to highlight quantitative relationships [30], while text should be used if the system wants to convey a precise value, such as the name of a city [54].
- *Contextual Information:* Under certain circumstances it may be appropriate for the system to consider properties of the user's environment when deciding how to best present information. As an example, for systems designed for use in vehicles, variables such as the current weather conditions and other relevant operating conditions

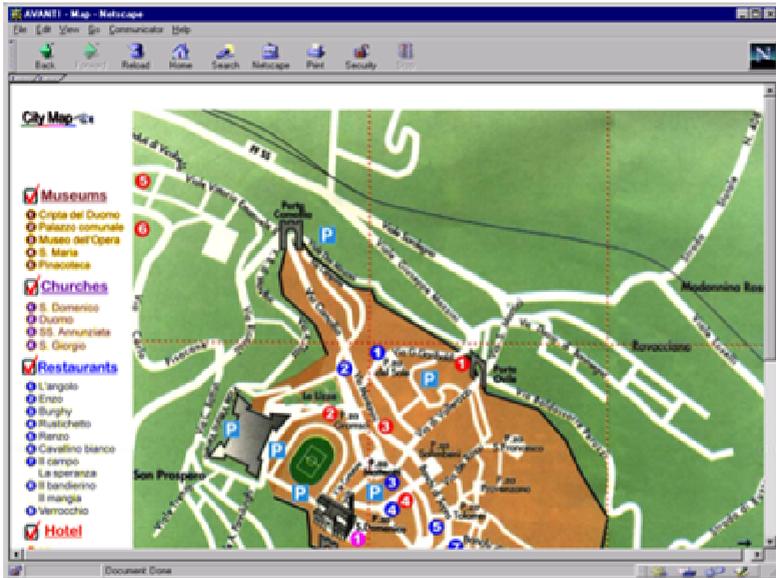


Fig. 13.5. The AVANTI system [16] with information presented graphically.



Fig. 13.6. The AVANTI system [16] with information presenting using text.

(e.g., speed and traffic) should influence both the appropriate media type and the quantity of information presented [12]. For instance, in situations where the weather is poor, or the user is driving in a high speed zone, designers are exploring the use of media such as haptics and speech to avoid placing additional burden on the already loaded visual channel (e.g., [4] and [12]).

- *Media Constraints*: When using multiple media, at times these media should cooperate to ensure the best overall presentation. For example, a mix of compatible media (e.g., text and speech) can increase the recallability of the information presented [54], and contrasting pieces of information should be presented using the same medium (e.g., [1] and [54]).
- *Limitations of Technical Resources*: This factor relates to the device on which the information will eventually be displayed and the possible media limitations of this device. Examples of resource limitations include the available bandwidth (e.g., large images should not be displayed if there is low bandwidth [16, 34]), the general availability of different media (e.g., whether or not speech is available [45]) or the available screen real estate.

**Example Applications.** We now describe some examples of systems that aim to perform some sort of media adaptation in an adaptive hypermedia system or adaptive Web site. The first two examples are forms of tourist information systems. The remaining examples are: a mobile navigation system, a learning environment and a pointer to a more general framework for generating multimedia presentations on the Web.

The AVANTI system [16] adapts the media through which the relevant information is presented to the user according to 1) accessibility issues, for example, using spoken language for visually-impaired users, and 2) resources issues, such as not presenting too many graphics in low-bandwidth situations. Fig. 13.5 and Fig. 13.6 illustrate two example versions of a page containing information on a city map: one for sighted users and another for visually-impaired users. For visually-impaired users, the map is described using text (which could presumably be read by a speech synthesizer) (Fig. 13.6). Alternatively, a regular graphical map is displayed for sighted users (Fig. 13.5).

The MASTROCARONTE system [12], an in-car tourist information system, adapts the medium through which recommendations are made to the user by considering 1) the user-specific factors, such as preferences and the user's current level of fatigue (estimated based on the time of day and the length of the current trip), and 2) contextual factors, such as speed and traffic volume. For example, if the user is requesting restaurant recommendations, the system determines the current level of risk as indicated by the contextual factors (e.g., traffic or visibility) and the user's current level of fatigue to decide a) whether to present the recommendations visually or using speech, and b) how many recommendations to present at one time. If the contextual factors indicate a high-level of risk (e.g., high traffic or low visibility) *or* if the user is fatigued, the system will elect to present the recommendations visually, and will present only one recommendation at a time (requiring the user to press a button to retrieve additional recommendations). If there is both a high-level of risk *and* the user is fatigued, the system will present the recommendations using speech. If neither condition holds, the system will present several recommendations at once and will present them visually.

In [34], the authors describe guidelines to be used by a mobile navigation system to decide which medium to use when presenting route information. These guidelines take into account technical resource limitations (the last factor described in the previous section), as well as the user's context and cognitive resources. Route descriptions using text and speech are appropriate only if the system knows the user's location and are useful when the amount of available bandwidth is small (since no images have to be sent to the mobile device). Speech is preferred to text in this scenario if the user is also experiencing a high level of cognitive load since it does not require visual attention. Graphical route instructions do not require location information (in their system), but require more cognitive and technical resources.

The CUMAPH adaptive hypermedia environment [21] adapts hypermedia documents according to a user profile that describes the user's cognitive abilities, such as the user's ability to explore visual and spatial information, and the user's visual and auditory memory. A description of how the profile is acquired can be found in [22]. The environment has so far been applied to an instructional domain involving a course about the human brain. As an example, when adapting a Web page on the topic of "memory", the system uses the optimization approach described in the next section to decide whether to display the content of two relevant sub items ("definition" and "mechanism") using text, graphics, sound or a combination of two of the three media.

Finally, the hypermedia document formatter described in [45] includes capabilities to adapt media presentation to device characteristics and user preferences. The paper does not illustrate an application of the framework. Instead, the authors focus primarily on the architecture necessary to realize multimedia presentations on the Web.

**General Approaches.** The most common techniques for media adaptation fall into 2 general categories: 1) *rule-based* or *planning* approaches and 2) *optimization* approaches. We will now describe each type of technique. Once again, we extend our discussion to include systems that are not adaptive Web-based systems since we feel that the techniques could certainly be generalized to adaptive Web applications.

*Rule-Based Approach.* The vast majority of systems that perform media adaptation do so using rules that describe how to best convey the target information given subsets of the factors described in earlier in the section. Examples of such systems include [2], [3], [15], [16], [19], [23], [37] and [45].

Andre [1] discusses key differences among several of the above systems. A primary difference is how integrated the content selection process (as discussed in section 13.2.2) is with the media allocation process (i.e., determining which medium to use for a given information element). The systems also differ in the type and amount of communication among the components in charge of realizing the media-specific information (to ensure a coherent and feasible overall result). To illustrate these differences and to provide a more concrete understanding of how media allocation using rules can work, we will now elaborate on two examples: the work by Arens *et al.* [3] and the WIP system [2].

Arens *et al.* [3] describe a system that can adapt the media based on characteristics of the information to be conveyed, media constraints, the user's interests and abilities, and the overall goals of the information presentation (referred to as the "presenter's

goals”). The system begins by selecting and structuring the content (using NLG techniques similar those discussed in section 13.2.2), which produces a tree that represents the discourse structure. The system then applies media allocation rules to the discourse structure to obtain a presentation structure. It applies these rules by traversing the tree in a bottom-up manner. As the rule-application process moves up the tree, earlier media decisions can be reconsidered based on information about the more global structure of the presentation, which is only visible at the higher-level nodes. Such a situation could occur if, for example, content item A is assigned to be presented using text and content item B graphics, but the higher-level discourse goal is to compare the two items. With this higher-level goal, both items should be presented using the same medium. Once the presentation structure is complete, each element is then sent to the appropriate generator, and all of the results are sent to a final layout specialist, which decides how to arrange them on the target display.

The WIP system, described in [2], interleaves content planning and media assignment. WIP focuses primarily on adapting the media based on characteristics of the information to be displayed, but the authors indicate that their technique could be extended to take additional factors into consideration. Generating the presentation involves a task-decomposition schema-based approach similar to what is often used for content structuring in NLG (see section 13.2.2). To permit media allocation, WIP’s schemas (referred to as “presentation strategies”) have an additional slot for the medium through which the information should be conveyed. Some schemas specify the appropriate medium, other schemas leave this slot open, to be filled later in the planning process. Using the schemas, the system engages in a top-down planning process, starting from the high-level presentation goal. When more than one schema is applicable, but they have different media assignments, the system selects which schema to use based on meta-rules relating to how well the media accomplish the current presentation goal. As soon as a media slot is filled, the given presentation goal is sent to the generator in charge of realizing information through that medium. This has the advantage that a given presentation goal can be refined by a generator, if it is not able to fulfill the request. Such a situation may occur if, for example, a generator is asked to display a piece of information using an image, but doing so would require an image that is larger than the amount of remaining screen real estate.

Andre [1] argues for content selection/structuring to be interleaved with media allocation because it allows the media allocation process to inform the discourse structure. However, if one does not want media selection to influence content selection/structuring, or wants to use an existing content selection/structuring process as is, it makes sense to employ the sequential approach. Arens *et al.* [3] also argue for the simplicity of the sequential approach.

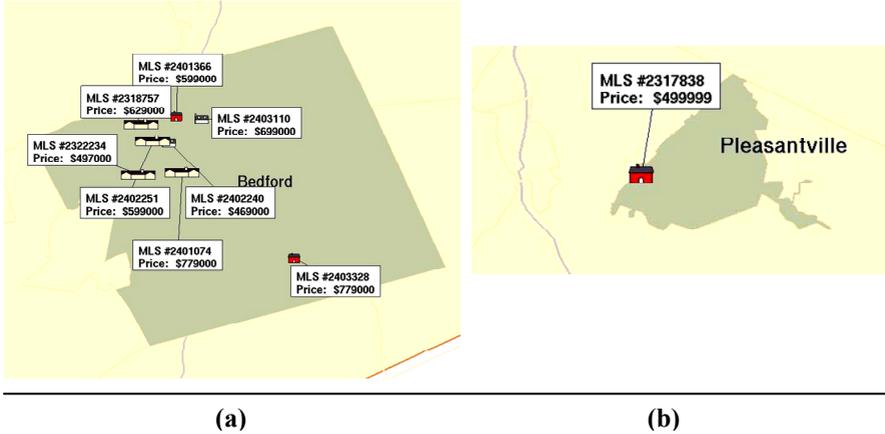
*Optimization Approach.* An alternative to a rule-based approach is to formulate the media adaptation process as an optimization problem. That is, given information on the relevant factors, the goal is to find the media combination that produces the best overall result. Examples of systems that follow this approach are the RIA system [54] and the CUMAPH system [21]. As discussed in section 13.2.2, RIA optimizes a set of feature-based metrics to perform tailored content selection. RIA follows a similar optimization approach to adapt the media once the relevant information elements have been selected. The authors present a comprehensive list of feature-based metrics that

**U2** *Speech*: How much are those?

**U3** *Speech*: How much is this one?

*Gesture*: Point to a house on the screen

**RIA** The prices are shown on the screen. **RIA** The asking price is \$499,000.



**Fig. 13.7.** Example of media adaptation in the RIA system [54]

they use to find the optimal media allocation. These metrics fall into two categories: 1) media selection metrics that indicate which media best suits a particular content item (based on user preferences, properties of the user's task, and properties of the content items themselves) and 2) presentation coordination metrics that describe the level of coordination among the different media. These metrics are combined in an overall objective function and the media allocation process then becomes the task of maximizing this objective function.

Fig. 13.7 shows the RIA system responding to two similar user requests (U2 and U3) with different media allocation decisions. In (a), the user (U2) is requesting the prices of a number of houses, whereas in (b) the user (U3) is requesting the price of a particular house. We use this example to show how the system's optimization procedure is able to deal with three metrics: the suitability of the information to the media, the desire to increase recallability, and the desire to maintain presentation consistency (the first is a media selection metric and the latter are two presentation coordination metrics). In (a), because of the high-volume of data, the price information should be displayed visually and speech can be used only to focus the user's attention to the screen. In contrast, the low data volume in (b) allows the system to display the price information visually and also to mention the price using speech, which increases recallability. In both examples, because house location, city name and city boundary are considered to be inter-related, all three are presented visually.

The CUMAPH system [21] follows a similar, but scaled-down approach. They have 2 metrics: one assigns the highest value to the media combination that best fits the user profile; the other favors combining multiple media (e.g., using both graphics and text as opposed to just graphics). The system generates all possible combinations of media assignments to information items and picks the one whose sum of the two

metrics is the highest. This type of exhaustive generation and testing is likely only possible when the number of combinations is small.

Both of these systems perform media allocation after the content has been selected. Future work in the area could involve investigating ways to use the optimization approach to interleave the two processes.

*Comparison of Approaches.* According to [53] and [54], the advantages of the optimization approach are: 1) it does not require a large set of rules and/or plans to be authored, 2) it allows the system to handle issues with conflicting or interdependent factors without a large amount of communication among different system components, 3) it is more easily extended, and 4) it is more easily transferred to different domains. A downside of the optimization approach is that there do appear to be cases where the system is required to repair the media allocation after the optimization process is complete [54]. In the case of [54], these situations are detected and repaired by relying on rules.

We are not aware of any evaluations that have directly compared the rule-based and the optimization approaches. In fact, there has been very little evaluation of systems that perform media adaptation. One exception can be found in [54], where RIA's multimedia presentations were evaluated using the human-designer approach (described in Chapter CH6 [18]). The results of the experiment were promising. Out of 50 test cases, RIA was rated (or co-rated) highest in 17 cases. In the 28 of the remaining cases, the difference between the winner and RIA was minor. For the other 5 cases, the authors felt that the problem was due to content selection rather than media selection. Thus, in the majority of cases (45 out of 50), RIA was able to perform as well, almost as well, or even better than a human interface designer.

## 13.4 Conclusions

In this chapter we discussed techniques for providing adaptive content presentation for the Web. We structured the discussion by first introducing techniques to select and coherently structure the content deemed to be most relevant for the current interaction context, and then by illustrating approaches to further adapt the selected content by tailoring the way in which it is actually presented to the user. In particular, we have described techniques to adapt the presentation based on content relevance, as well as techniques to adapt the type of media for optimal content delivery.

Adaptive presentation of Web content can complement several of the other types of adaptations presented in this book. For instance, it can be integrated with both adaptive navigation support and adaptive search to further adapt the content of the Web pages returned by these two techniques to a user's needs; or it can be coupled with recommender systems to provide tailored descriptions of the recommended items. Although there has been some work on exploiting the integration of adaptive presentation with other types of adaptations (e.g., [10]) we see this as an area that still yields a high potential for innovative research in the adaptive Web. Another area that calls for strong research efforts is the validation of techniques for adaptive presentation. In this chapter we have discussed the results of some evaluations; however, the actual effectiveness of the many of the techniques is still largely open to investigation.

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