

ture of the user's notes which is used to provide task-relevant recommendations.

When a user begins a new task, they create a new workspace which provides a blank space to record new notes. The workspace allows users to drag-and-drop information from the browser into their notes as shown in Figure 2. Users can drag text, images, links, complex combinations of these types. Users can also record entire web pages by dragging the location bar's link icon to the InsightFinder. In addition to adding new information, users can manipulate and organize existing objects within the InsightFinder to clarify their overall notes as their task evolves. This includes tools for folder creation and manipulation, as well as deleting, moving, editing, and revisiting existing notes.

The InsightFinder provides more than basic note taking capabilities. Users can save their notes and return to them in future sessions. In addition, users can maintain unique multi-session workspaces for each user task. This is especially critical because users often engage in several interleaved sensemaking tasks over the course of several web browsing sessions. Task-specific workspaces (e.g., one for "Trip to New York" and another for "Investment Research") allow the InsightFinder to provide more accurate task-specific page unit recommendations.

Underneath the graphical display of the user's collection of notes is a graph-based data structure, called the *context model*. Each task t has its own model, C_t , that mirrors the workspace's visual presentation and augments it with additional information required for the page unit recommendation algorithm. Every change to a user's notes is automatically propagated to the underlying context graph.

2.2 Providing Recommendations

The InsightFinder is designed to dynamically recommend the most task-relevant content on each web page visited by a user. This is done by comparing the content on each visited page to the information stored in the user's notes. The results are presented visually to the user to reduce the time spent searching through pages to find relevant information.

The recommendation algorithm, expressed in Equation 1, is re-evaluated every time a user navigates to a new web page. The InsightFinder employs a structure-based algorithm to segment each page P_i into a set of individual page units, $p_{i,k}$, with the goal of creating units that contain semantically consistent data (i.e., the content within a single page unit should share a common topic).

$$p_{i,k} = \text{Rank}(\text{Segment}(P_i), C_t) \quad (1)$$

A relevance algorithm, *Rank*, then compares the information stored in the active context model C_t with the content of each of the page units extracted from the browser's current page, P . The *Rank* algorithm returns a vector of page units ranked by each one's computed degree of relevance to C_t . Our prototype uses a bag-of-words representation for page units and objects in C_t , and employs traditional text-oriented measures (e.g., the Jaccard coefficient and pointwise mutual information) within the *Rank* algorithm.

The most relevant page units in $p_{i,k}$ are recommended to the user through the InsightFinder's sidebar. The lower portion of the interface presents a sorted histogram providing an intuitive display of the degree of relevance for each of the recommended units. This interface is shown in Figure 3(a). A user's click on any item in the list causes the

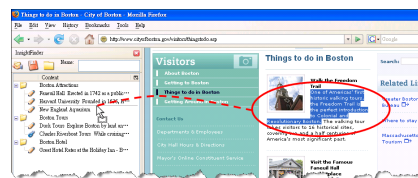


Figure 2: Users record notes by dragging content fragments (links, images, text, or entire pages) from the browser to folders in the InsightFinder.

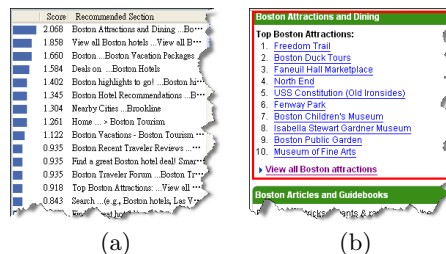


Figure 3: (a) The InsightFinder's recommendations are presented as a ranked list. A histogram is included to visually indicate the relative value of each unit's relevance to the context model. (b) Clicking on any item in the list automatically scrolls the web page and highlights the page unit in a red rectangle.

browser to automatically scroll to display the selected page unit. In addition, as shown in Figure 3(b), the selected unit is highlighted within a red box to clearly indicate where the recommended content can be found.

Relevance computation is an important part of the InsightFinder's ability to support sensemaking tasks. It helps users "connect the dots" by highlighting potentially relevant connections between their notes and the information currently on display within their browser. This feature can prove especially useful in quickly uncovering either intended or serendipitous connections which a user would otherwise overlook or obtain only by tediously analyzing the entire page.

3. EVALUATION

While the InsightFinder is only an early prototype, our initial evaluations show that it is effective in reducing the time required to locate relevant information within a web page. In laboratory experiments, we asked 10 users to perform four information location tasks using the Firefox browser. In all four tasks, users with access to the InsightFinder exhibited a statistically significant reduction ($p < 0.01$) in the time required to perform their task. These early results motivate our continuing work in this direction.

4. REFERENCES

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