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Video Browsing Interfaces and Applications: A Review

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ABSTRACT

This paper presents a comprehensive review of the state of the art on video browsing and retrieval systems, with special emphasis on interfaces and applications. It was motivated by the significant increase in activity (e.g., storage, retrieval, sharing) employing video data in the past decade, both for personal and professional use. The ever-growing amount of video contents available for human consumption and the inherent characteristics of video data – which, if presented in its raw format, is rather unwieldy and costly – have become driving forces for the development of more effective solutions to present video contents and allow rich user interaction. As a result, there are many contemporary research efforts towards developing better video browsing solutions, which are summarized in this paper. Our survey reviews more than 40 different video browsing and retrieval interfaces and classifies them into three groups: applications that use video player-like interaction, video retrieval applications, and browsing solutions based on video surrogates. For each category, we present a summary of existing work, highlighting the technical aspects of each solution, and comparing them against each other.

Keywords: video browsing, video navigation, review, survey

1. INTRODUCTION

The main research motivation in interactive information retrieval is to support users in their information seeking process. Salton\textsuperscript{1} defines a classical information seeking model as follows. Triggered by an information need, users start formulating a search query, inspect retrieval results and, if needed, re-formulate the query until they are satisfied with the retrieval result. Belkin et al.\textsuperscript{2} extend this model further by distinguishing between querying/searching for results, usually by triggering a new search query, and browsing/navigating through the already retrieved results. However, users of information retrieval systems have very often only a very fuzzy understanding of how to find the information they are looking for. According to Spink et al.,\textsuperscript{3} users are often uncertain of their information need and hence have problems finding a starting point for their information seeking task. And even if users know exactly what they are intending to retrieve, formulating a “good” search query can be a challenging task. This problem even exacerbates when dealing with multimedia data. Formulating search query hence plays an important role in this task. Graphical user interfaces serve here as a mediator between the available data corpus and the user. It is the retrieval systems’ interface which will provide users facilities to formulate search queries and/or to dig into the available data. Hearst\textsuperscript{4} outlines various conditions that dominate the design of state-of-the-art search interfaces. First of all, the process of searching is a mean toward satisfying an information need. Interfaces should therefore avoid being intrusive, since this could disturb the users in their seeking process. Moreover, satisfying an information need is already a mentally intensive task. Consequently, the interface should not distract the users, but rather support them in their assessment of the search results. Especially in the WWW domain, search interfaces are not used by high expertise librarians only but also by the general public. Therefore, user interfaces have to be intuitive to use by a diverse group of potential users. Consequently, widely used web search interfaces such as Google, Bing or Yahoo! consist of very simple interfaces, mainly consisting of a keyword search box and results being displayed in a vertical list.

Considering the success of above mentioned web search engines, it is not premature to assume that these interfaces effectively handle the interaction between the user and the underlying text retrieval engine. However, text search engines are rather simple in comparison to their counterparts in the video retrieval domain. Therefore,
Jaimes et al.\textsuperscript{5} argue that this additional complexity introduces further challenges in the design of video retrieval interfaces.

The first challenge is how users shall be assisted in formulating a search query. Snoek et al.\textsuperscript{6} identified three query formulation paradigms in the video retrieval domain: Query-by-textual-keyword, Query-by-visual-example and Query-by-concept. Query-by-textual-keyword has been largely studied in the last decades and thus is a well-established search paradigm. Visual queries arise from content-based image retrieval systems. Users can provide an example image, select a set of colors from a color palette or sketch images and the underlying retrieval engine retrieves visually similar images. Query-by-concept includes the allocation of low-level features to high-level concepts. Basic examples are concepts such as outdoor vs. indoor,\textsuperscript{7} and cityscape vs. landscape\textsuperscript{8} which can be identified based on visual features. Concepts can be used to filter search results, e.g. by displaying only the results which depict a landscape. Video retrieval interfaces need to be provided with corresponding query formulation possibilities in order to support these paradigms. Another challenge is how videos shall be visualized allowing the user an easy understanding of the content. In the text retrieval domain, short summaries, referred to as snippets, are usually displayed which allow the users of the system to judge the content of the retrieved document. Multiple research (e.g.\textsuperscript{9,10}) indicate that such snippets are most informative when they show the search terms in their corresponding context. Considering the different nature of video documents and query options, identifying representative video snippets is a challenging research problem. Moreover, another challenge is how users can be assisted in browsing the retrieved video documents. Systems are required which enable users to interactively explore the content of a video in order to get knowledge about its content.

In this paper, we survey representative state-of-the-art video browsing and exploration interfaces. While research on video browsing was very active already in the 1990’s (e.g.\textsuperscript{11–26}) in this paper we focus on video browsing approaches that have been presented in the literature during the last 10 years. Many systems reviewed in this paper have been evaluated within TRECVid,\textsuperscript{27} a series of benchmarking workshops aiming to improve content based video retrieval techniques. The paper is structured as follows. In Section 2, we review video browsing applications that rely on interaction similar to classical video players. Section 3 introduces applications that allow users to explore video corpus using visual keyframes. Section 4 surveys video browsing applications that visualize video content in unconventional ways. The paper concludes in Section 5.

## 2. VIDEO BROWSING APPLICATIONS USING VIDEOPLAYER-LIKE INTERACTION

Common video players use simple interaction means to navigate through the content of a video. However, although these interaction means are often employed for the task of searching, they are mostly unsatisfying. Therefore, many efforts have been made to extend the simple video players’ interaction model with more powerful means for content-based search. In this chapter we review such video browsing applications which can be characterized as ”extended video players”.

One of the early efforts in this direction has been done by Li et al.\textsuperscript{28} in 2000. They developed two different versions of a video browser, a \textit{basic browser} and an \textit{enhanced browser}, and compared both versions in a user study with 30 participants. The basic browser included basic controls, those which are typically provided by video players, such like play, pause, fast-forward, seeker-bar, etc. The enhanced browser provided several additional features:

- The \textit{time compression (TC)} function increases/decreases playback speed from 50% to 250% by always preserving the audio pitch.
- The \textit{pause removal} function removes segments which seem to be silence or pause, according to the audio channel.
- The \textit{Table of Content (TOC)} feature is a list of textual entries (e.g. for ”classroom” videos, generated from the corresponding slides).
- A \textit{visual index} containing keyframes of all the shots.
Jump backward or jump forward by 5 or 10 seconds; jump to next note; jump to next slide transition ("classroom" videos) or shot change (shot boundary seek)

In their evaluation, they showed that users of the enhanced browser rated TC and TOC as most useful features while the shot seek feature has been used most often. Moreover, their evaluation showed that participants spent considerably less time on watching videos with default playback speed when using the enhanced browser. It also revealed that the fast-forward feature of the basic browser has been used significantly less than the seeker-bar. For the classroom and the news video fast-forward has almost never been used. However, for the sports category (baseball video) the average number of fast-forward usage heavily increased for both the basic and the enhanced browser because it allowed higher speed-up than TC. Participants agreed that especially in the sports and news category, having enhanced browsing features would be of great benefit and affect the way they watch television.

Barbieri et al.\textsuperscript{29} presented the concept of the \textit{Color Browser}, where the background of a usual seeker-bar is enhanced by vertical color lines, representing information about the content. As information to be presented in the vertical lines they used (i) the dominant color of each corresponding frame (Figure 1) and (ii) the volume of the audio track. For the dominant colors a smoothening filter is applied to filter out successive heavily changing color values (see Figure 1). As there is not enough space to display a vertical line for every frame in the background of a seeker-bar, they proposed to use two seeker-bars. The first one acts as a fast navigation mean with different time scale for every video sequence and the second acts as a time-related zoom using the same time scale for every video sequence. They argued that the fixed time scale of the zoomed seeker-bar would enable a user to "learn to recognize patterns of colors within different programs".

\begin{figure}[h]
\centering
\includegraphics[width=0.5\textwidth]{color_browser.png}
\caption{Visualization of the \textit{ColorBrowser} without(above) and with(below) a smoothening filter\textsuperscript{29}}
\end{figure}

Tang et al.\textsuperscript{30} presented the \textit{NewsEye}, an application for improved news story browsing. With unsupervised fuzzy c-means clustering the content is first segmented into shots. Then, the shots are grouped together in order to form several different news stories. For that purpose they use a graph-theoretical cluster analysis algorithm to identify all shots that show an anchorperson. Furthermore, they also use OCR to detect caption text in the frames of a news story. Their VideoPlayer-like interface contains an own panel showing keyframes of all the shots in the current news story as well as the detected caption text. Their application also provides a keyword-based search function for the caption text.

Divakaran et al.\textsuperscript{31} proposed a video summarization method that can also be used for video browsing. Their approach take advantage of information extracted from the compressed domain of the video and it is based on the hypothesis that the intensity of motion activity is a measure of the summarizability. To skip over parts of the video with low motion activity, the playback rate is adjusted dynamically. They also analyze the audio channel in order to detect speaker changes and to generate a list of included topics.

In a further work Divakaran and Peker\textsuperscript{32} propose to use \textit{Adaptive Fast Playback} (AFP) for the purpose of quickly skimming through a video sequence. The AFP approach is used accordingly to the \textit{level of complexity} of a particular scene and the capabilities of the human visual system. The level of complexity is determined based on the amount of motion and spatial-temporal complexity of a scene. Thus, scenes with low complexity are played faster while scenes with high complexity are played at lower speed.

Liu et al.\textsuperscript{33} presented a news video browsing system called \textit{NewsBR}, which is very similar to the \textit{NewsEye} system.\textsuperscript{30} It performs story segmentation and caption text extraction. The story segmentation uses a shot detection method based on $\chi^2$ histogram matching and silence clip detection. For caption text extraction they
classify frames into *topic-caption frames* and *non topic-caption frames*. To topic-caption frames, which are those that contain text of a news topic, a horizontal and vertical sobel filter is applied before an OCR library is used to detect the text. Their interface shows a TOC (in combination with a keyframe preview) according to the story segmentation which can be used as navigation means. It also provides a keyword-based search on the extracted caption text.

Moraveji\textsuperscript{34} proposed to assign unique and visually distinctive colors to particular content features, such as persons, faces, vehicles, etc. These colors can be further used for visualization in a timeline which shows "the most relevant" feature for a particular segment of frames (see Fig. 2). When the mouse is moved over a particular segment, some additional information - such as the concept/feature represented by the color - is displayed below. A click on a color bar in the timeline will start video playback from the corresponding time position. The work of Moraveji is similar to the work of Barbieri et al.,\textsuperscript{29} as it is based on the idea of enhancing the timeline (or background of a seeker-bar) with content information.

To overcome the limitations of usual seeker-bars in standard video players, Hürtst et al. proposed the Zoom-Slider interface.\textsuperscript{35,36} Instead of a common seeker-bar the entire player window is used as a hidden seeker-bar with different stages of granularity in a linear way (see Figure 3). When the user clicks on any position in the player window, a seeker-bar for moving backward or forward appears. The granularity of that seeker-bar is dependent on the vertical position of the mouse in relation to the entire height of the player window. When the mouse is moved in vertical direction the scaling of the seeker-bar changes in a linear way. The finest granularity is used at the top of the window and the coarsest granularity is used at the bottom of the window. Therefore, a user can zoom-in or zoom-out the scaling of the seeker-bar by selecting different vertical mouse positions.

The concept of the ZoomSlider interface has been extended in\textsuperscript{37} to additionally provide similar mechanisms for changing the playback speed of the video. The right vertical part of the player window is used to change the
playback speed where the slowest speed is assigned to the top and the highest speed is assigned to the bottom of the player window. The user can select any playback speed in a linear fashion based on the vertical mouse position. The same manner is used for backward playback at the left vertical part of the window. In\textsuperscript{38,39} the idea has been further adapted for mobile devices, where the entire screen is used for video playback containing "virtual" seeker-bars in the same way.

Divakaran and Otsuka\textsuperscript{40} argued that "Current Personal Video Recorders can store hundreds of hours of content and the future promises even greater storage capacity. Manual navigation through such large volumes of content would be tedious if not infeasible." Therefore, they presented a content-based feature visualization concept (Figure 4) which is based on classification of audio segments into several different categories (e.g. Speech, Applause, Cheering, etc.). An importance level is calculated according to these categories and plotted in a two-dimensional graph, which can be shown as a timeline overlay onto the original content. The user can set an importance level threshold (yellow line in the figure), which is used by the system to filter out all the content having a lower importance level. In other words, a "Highlight Search" function is available to the user. They evaluated their concept with several sports videos in a user-study which has shown that users like the importance level plot due to its flexibility, even if the visualization results in mistakes. The concept has been integrated into a Personal Video Recorder product sold by Mitsubishi Electric in Japan.

![Figure 4. A video browsing enhanced personal video recorder\textsuperscript{40}](image)

An interesting approach for video browsing by direct manipulation has been presented by Dragicevic et al.\textsuperscript{41} in 2008. As a complement to the seeker-bar they propose relative flow dragging (Figure 5), which is a technique to move forward and backward in a video by direct mouse manipulation (i.e. dragging) of content objects. They use an optical flow estimation algorithm based on SIFT\textsuperscript{42} salient feature points of two consecutive frames. A user study has been conducted and it has shown that relative flow dragging can significantly outperform the seeker-bar on specific search tasks.

A system very similar to that of Dragicevic et al. has already been proposed by Kimber et al. in 2007.\textsuperscript{43} In similarity, their system shows motion trails of objects in a scene, based on foreground/background segmentation and object tracking, and allows dragging an object along a trail with the mouse. For the application in a floor surveillance video they additionally show the corresponding floor plan including motion trails.

Chen et al. presented the EmoPlayer,\textsuperscript{44} a video player which can visualize affective annotations. In particular, different emotions of actors and actresses - angry, fear, sad, happy, and neutral - can be visualized for a selected character in a video, based on a manually annotated XML file. The emotions are shown in a color-coded bar directly above the usual seeker-bar. Different colors are used for different emotions (see Figure 6). If a character is not present in a specific scene the bar shows no color (i.e. white) for the corresponding segment. Therefore, a user can simply identify in which segments a particular character is present and which emotion the character expresses.
In 2008, Yang et al. proposed the *Smart Video Player* to facilitate browsing and seeking in videos. It provides a filmstrip view in the bottom part of the screen, which shows keyframes of the shots of the video (see Figure 7). The user can set the level of detail for that view and, thus, extend or reduce the number of shots displayed within the filmstrip. The Smart Video Player does also contain a recommendation function to present a list of other similar videos to the user.

In 2002, a similar technique has been presented by Drucker et al. with the *SmartSkip* interface for consumer devices (e.g. VCRs). They propose a thumbnail view at the bottom of the screen (see Figure 8) which can be used
to skip over less interesting parts of the video. These thumbnails have been uniformly selected from the content of the video although they experimented with a shot-based view as well. The shot-based view, however, has been omitted after user tests. The reason was that for communicating the actual time between shots the spatial layout has been changed to a non-uniform manner, which users disliked. The level-of-detail of the thumbnail view can be configured by users ranging from 10 seconds all the way up to 8 minutes.

Huayong et al.\textsuperscript{47} combine video event analysis with textual indices for the SportBR video browser, that can be used for browsing soccer videos. In particular, they use the color layout of example images of penalty kicks, free kicks, and corner kicks and search for similar scenes in the video. In order to improve the accuracy of the event detection, speech analysis (detection of some specific words) is performed. Moreover, they use an OCR algorithm to detect text that appears in the frames within detected events. The interface of their application allows (1) improved navigation within the video based on the detected events and (2) keyword search based on speech and text detection.

Vakkalanka et al.\textsuperscript{48} presented the NVIBRS, a news video indexing, browsing and retrieval system. Their system performs shot detection and news story segmentation based on localization of anchorperson frames. To detect anchorperson frames they first classify all frames into high motion and low motion. On low motion frames they apply a face detection method based on a Gaussian Mixture Model for skin-color detection. When a face has been detected the location of the eyes and the mouth is estimated and features are extracted from that regions. The feature vectors are used as input for an anchorperson classifier working with autoassociative neural network models. The interface of their application provides a tree view of all detected news story units in a video.
and shows keyframes of the currently selected story as a navigation means. It also allows a user to perform a textual query by specifying the desired video category as the news content is categorized into a few categories.

Rehatscheck et al.\textsuperscript{49} and Bailer et al.\textsuperscript{50} presented the \textit{Semantic Video Annotation Tool (SVAT)}, a tool that is basically intended to be used for video annotation (see Figure 9). However, in order to improve navigation within a single video for faster annotation they developed several advanced navigation functions. In particular, they provide a video player like component (1) in combination with a temporal visualization of shot boundaries, keyframes, stripe images, and motion events (pan, zoom, etc.) as navigation means (2). Their interface also includes a shot list (3), a list for selected keyframes (4), and an annotation view (5) to add textual information to shots and keyframes. Moreover, the tool includes a SIFT-based automatic similarity search function which can be used to find similar content in the video according to a user-defined region of interest.

A similar tool for the explicit purpose of video browsing has been presented by Schoeffmann et al.\textsuperscript{51} Their \textit{Video Explorer} uses the concept of Interactive Navigation Summaries (INSs) in order to help a user at the task of navigation through a video. INSs can effectively visualize several time-related information. As shown in Figure 10, the Video Explorer consists of a video player like component (1) and a few INSs (2 and 3) that act as an alternative to the common seeker-bar. (2) shows the \textit{dominant color INS} and (3) shows the \textit{motion layout INS}. While the dominant color INS\textsuperscript{52} visualizes the temporal flow of a dominant colors, the motion layout INS\textsuperscript{53} visualizes the temporal flow of motion characteristics. More precisely, for the second INS motion vectors of H.264/AVC compressed video files are extracted, classified by direction and intensity, and visualized in an HSV color representation. A Hue circle of the HSV color space is shown at (4) in order to give the user a hint which color is used to visualize a particular direction (e.g. blue for motion down, yellow for motion up, red for motion right up and so on). The visualization shows both how much motion to a specific direction every frame contains (amount of a specific color (H) in a vertical line) and how fast this motion is (intensity (V) of the color). For a specific scene this yields to a certain \textit{motion pattern} that can help users to interactively detect similar scenes in the video, as they appear with similar motion patterns in the visualization. Figure 10 shows an example of a ski-jumping video where jump-offs of competitors are visualized as greenish V-like patterns. In order to preserve the browsing context their model of an INS contains an \textit{overview visualization}, including a \textit{zoom-window}, and a \textit{detailed visualization}. While the overview visualization represents the entire video in low
quality, the detailed visualization (located directly below to the overview) shows all the details of a particular segment. The zoom-window (shown as a red box) determines the position and duration of this segment to be shown in the detailed visualization of the corresponding INS.

Figure 10. Video Browsing with the Video Explorer

Cheng et al. proposed the SmartPlayer (see Figure 11) for browsing the content of a video. In addition to manually changing the playback speed it provides an automatic playback speed adaptation according to scene complexity, which is computed through motion complexity analysis for every shot. The player has been designed in accordance with the “scenic car driving” metaphor, where a driver slows down at interesting areas and speeds up through unexciting areas. The SmartPlayer also learns the users’ preferences of playback speed for specific type of video content.

Figure 11. The SmartPlayer
Tables 1 and 2 give an overview of approaches reviewed in this section. All applications in this section and in the two subsequent ones have been structured by the following criteria:

- Support for *browsing* and *querying* the content?
- Is the application intended to be used with a *single video file* (1) or an *archive* (N)?
- What is the smallest *structuring unit* the content analysis and interaction is bound to?
- What is the *video content domain* the application is designed for?
- Which *content analysis* is used?
- How is the *content visualization/representation* and *user interaction* implemented?

<table>
<thead>
<tr>
<th>Browsing/Querying</th>
<th>Input Files</th>
<th>Struct. Unit</th>
<th>Video Dom.</th>
<th>Content Analysis</th>
<th>Visualization / Interaction</th>
</tr>
</thead>
<tbody>
<tr>
<td>Li et al.²⁸</td>
<td>yes/no</td>
<td>1</td>
<td>all</td>
<td>audio/speech analysis (pause removal), shot boundary detection, text recognition</td>
<td>Similar to a video player with speeded-up playback and navigation indices</td>
</tr>
<tr>
<td>Barbieri et al.²⁹ <em>(ColorBrowser)</em></td>
<td>yes/no</td>
<td>1</td>
<td>all</td>
<td>dominant color, audio-track volume</td>
<td>colored seeker-bar</td>
</tr>
<tr>
<td>Tang et al.³⁰ <em>(NewsEye)</em></td>
<td>yes/yes</td>
<td>1</td>
<td>story</td>
<td>news unsupervised clustering techniques for shot boundary detection and story segmentation, OCR</td>
<td>Similar to a video player with advanced navigation helps and caption text display/search</td>
</tr>
<tr>
<td>Divakaran³¹</td>
<td>yes/no</td>
<td>1</td>
<td>all</td>
<td>MPEG-7 motion activity</td>
<td>Adaptive fast playback</td>
</tr>
<tr>
<td>Peker et al.³²</td>
<td>yes/no</td>
<td>1</td>
<td>all</td>
<td>temporal frequency and spatio-temporal complexity based on DCT blockhistograms</td>
<td>Adaptive fast playback</td>
</tr>
<tr>
<td>Liu et al.³³ <em>(NewsBR)</em></td>
<td>yes/yes</td>
<td>1</td>
<td>story</td>
<td>news shot boundary detection ($\chi^2$), silence detection, sobel filtering, OCR</td>
<td>Similar to a video player with advanced navigation helps and caption text search</td>
</tr>
<tr>
<td>Moraveji et al.³⁴ <em>(Color Bars)</em></td>
<td>yes/no</td>
<td>1</td>
<td>all</td>
<td>text-based annotation</td>
<td>2D visualization through color bars as a seeker-bar</td>
</tr>
<tr>
<td>Hürt et al.³⁵-³⁶ <em>(Zoom Slider)</em></td>
<td>yes/no</td>
<td>1</td>
<td>(time)</td>
<td>not required</td>
<td>common video player</td>
</tr>
<tr>
<td>Divakaran³⁷*(PVR)*</td>
<td>yes/no</td>
<td>1</td>
<td>sports</td>
<td>audio volume analysis</td>
<td>2D audio volume plot as a seeker-bar</td>
</tr>
</tbody>
</table>

Table 1. Overview of VideoPlayer-like Video Browsing Applications
<table>
<thead>
<tr>
<th>Browsing/Querying</th>
<th>Input Files</th>
<th>Struct. Unit</th>
<th>Video Dom.</th>
<th>Content Analysis</th>
<th>Visualization / Interaction</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dragicevic et al.(^{41}) (dimP)</td>
<td>yes/no</td>
<td>frame</td>
<td>all</td>
<td>optical motion flow estimation with SIFT</td>
<td>Flow dragging based on optical flow estimation</td>
</tr>
<tr>
<td>Kimber et al.(^{43})</td>
<td>yes/no</td>
<td>frame</td>
<td>surveillance</td>
<td>foreground/background segmentation and object tracking</td>
<td>Flow dragging based on optical flow estimation</td>
</tr>
<tr>
<td>Chen et al.(^{44}) (EmoPlayer)</td>
<td>yes/no</td>
<td>scene</td>
<td>videos containing emotions</td>
<td>manual annotations</td>
<td>colored seeker-bar</td>
</tr>
<tr>
<td>Yang et al.(^{45}) (Smart Video Player)</td>
<td>yes/no</td>
<td>shot</td>
<td>all</td>
<td>shot boundary detection, annotation-based similarity analysis of shots</td>
<td>filmstrip view of keyframes based on a user-selected level of detail, recommendation function</td>
</tr>
<tr>
<td>Drucker et al.(^{46}) (SmartSkip)</td>
<td>yes/no</td>
<td>frame</td>
<td>all</td>
<td>not required</td>
<td>filmstrip view of keyframes based on a user-selected level of detail</td>
</tr>
<tr>
<td>Huayong et al.(^{47}) (SportBR)</td>
<td>yes/yes</td>
<td>frame</td>
<td>soccer</td>
<td>color layout, speech and text recognition</td>
<td>video player with additional features for navigation and text-based search</td>
</tr>
<tr>
<td>Vakkalanka et al.(^{48}) (NVIBRS)</td>
<td>yes/yes</td>
<td>story</td>
<td>news</td>
<td>shot boundary detection, motion analysis, face detection</td>
<td>news browsing by a tree of news story units</td>
</tr>
<tr>
<td>Rehatscheck et al.(^{49,50}) (SVAT)</td>
<td>yes/yes</td>
<td>frame</td>
<td>all</td>
<td>shot boundary detection, temporal view of stripe images, keyframes and motion events; content-based similarity search</td>
<td></td>
</tr>
<tr>
<td>Schoeffmann et al.(^{51–53}) (Video Explorer)</td>
<td>yes/no</td>
<td>frame</td>
<td>all</td>
<td>dominant color extraction, motion analysis (motion vector classification)</td>
<td>Interactive Navigation Summaries visualizing the temporal flow of dominant colors and motion characteristics</td>
</tr>
<tr>
<td>Cheng et al.(^{54}) (SmartPlayer)</td>
<td>yes/no</td>
<td>shot</td>
<td>all</td>
<td>shot boundary detection based on color histograms, optical flow analysis</td>
<td>&quot;scenic car driving&quot; representation and automatic playback speed adaptation</td>
</tr>
</tbody>
</table>

Table 2. Overview of VideoPlayer-like Video Browsing Applications (cont’d)
3. VIDEO BROWSING CONCEPTS IN VIDEO RETRIEVAL APPLICATIONS

While browsing videos using a video player like interaction scheme is useful in some scenarios, this approach cannot easily be adopted in interactive video retrieval. In contrast to video browsing, where users often just interactively browse through video files in order to explore their content, a video retrieval user wants to search certain scenes in a collection of videos. Such a user is typically expected to know quite exactly what she is looking for. Therefore, it is crucial to provide appropriate search functions for different types of queries. However, at least for the task of presenting the results to a query, a video retrieval application needs to consider video browsing concepts as well. Furthermore, video browsing mechanisms are often combined with video retrieval methods (e.g. in VAST MM\textsuperscript{55}) in order to serve all different types of users. Nowadays interactive web-based video retrieval is also getting more important as both retrieval giants Yahoo! and Google are working on their own video retrieval engine. In addition there are numerous video search engines such as www.truveo.com and www.blinkx.com that offer similar services. These online video platforms allow users to upload and share their own videos. The data set of such platforms grows extremely quickly and necessitates new ways that allow users to efficiently browse through a large collection of videos. Since a review on arising challenges in the multimedia retrieval domain is out of the scope of this paper, the interested reader is referred to Veltkamp et al.\textsuperscript{56} for further reading. In this section we focus on different interface designs used in video retrieval systems.

In one of the earlier efforts for supporting video retrieval, Arman et al.\textsuperscript{57} proposed to use the concept of keyframes (denoted as \textit{Rframes} in their paper), which are representative frames of shots, for chronological browsing the content of a video sequence. Their approach uses simple motion analysis to find shot-boundaries in a video sequence. For every shot a keyframe is selected by using shape and color analysis. In addition to chronological browsing of keyframes, their approach already allows selecting a keyframe and searching for other similar keyframes in the video sequence. For visualization of the results, they proposed to display good results in original size (e.g. 100%), somewhat similar results in smaller size (e.g. 33%) and bad results in an even smaller size (e.g. 5%). Several other papers have been published, which use keyframe based browsing of shots in a video sequence, usually by showing a page-based grid-like visualization of keyframes (this is also called \textit{Storyboard}).\textsuperscript{58–67} Some of them propose clustering of keyframes into a hierarchical structure.\textsuperscript{58, 60, 63, 65} Considering the large amount of systems that visualize search results in a storyboard view, this approach can be seen as the standard visualization method. In the remainder of this section, we survey few representative interfaces which rely on this visualization paradigm. An introduction on different paradigms is given by Christel.\textsuperscript{68}

First efforts to provide a digital library started in 1996. The researchers from the University of North Carolina at Chapel Hill indexed short video segments of videos and joined them with images and text and hyperlinks in a dynamic query user interface. Their project evolved since then so that now, digitalized video clips from multiple sources are combined to the Open Video Project.\textsuperscript{69} Figure 12 shows a screenshot of the actual interface. It allows to trigger a textual search by entering a query, denoted (1) in the screenshot and the possibility to browse through the collections. Results are listed based on their importance to the given search query, denoted (2) in the screenshot.

![Figure 12. Open Video Graphical User Interface (screenshot taken from online system)](image-url)
Deng and Manjunath\textsuperscript{70} introduce a system using low-level visual features for content-based search and retrieval. Their system is based on shots and uses automatic shot partitioning and low-level feature extraction from compressed and decompressed domain. More specifically, videos are indexed using 256 bin RGB color histograms, motion histograms computed from MPEG motion vectors, and Gabor texture information. By giving an example shot, their system is able to retrieve similar shots of a video according to the three mentioned low-level features. The user may change the weights of the similarity matching for each of the three features.

Komlodi et al.\textsuperscript{61, 71} revealed in their user study that keyframe-based approaches such as the storyboards are still the preferred methods for seeking, even if additional time is required to interact with the user interface (scroll bars) and for eye movements. Dynamic approaches such as slideshows often display the content with a fixed frame rate and don’t allow the user to adjust it.

An alternative approach to the linear storyboard navigation is to present keyframes in a layered/hierarchical manner.\textsuperscript{65} At the top level a single keyframe represents the entire video, whereas the number of keyframes is increased at each level. If additional semantic information was extracted (e.g., an importance score), keyframes may be displayed in different sizes, drawing the user’s attention to important keyframes in the first place.\textsuperscript{59, 64} These scores can also be applied to dynamic approaches to adjust the playback speed and skipping unimportant scenes.

From 1998 to 2001, INRIA and Alcatel Alstom Research (AAR) developed the VideoPrep system which allows automatic shot-, keyframe-, object-, and scene-segmentation. The corresponding viewer VideoClic is able to provide direct linking between e.g. the same objects found on different temporal places. Some details about that work can be found in,\textsuperscript{72} pp. 20-23.

With the CueVideo project, Srinivasan et al.\textsuperscript{73} have presented a browsing interface which allows several visualizations of the video content. Their system is based on shots and consists of visual content presentation, aural content presentation, and technical statistics. Visual content presentation comprises (1) a storyboard where for each shot a keyframe is presented, and a (2) motion storyboard where for each shot an animated image is presented. The audio view shows a classification of the audio tracks into the categories music, speech, and interesting audio events. In a user study they found out that the most popular view was the storyboard view, which is a similar result as already found by Komlodi et al.\textsuperscript{61, 71} Users criticized the miss of the ‘top 10 keyframes’ and the bad scaling of the storyboard for long videos, but found it helpful (for content comprehension) to have different views.

Heesch et al.\textsuperscript{74} presented a tool for video retrieval and video browsing (Fig. 13), which they have used for TRECVID.\textsuperscript{75} The tool allows searching and browsing a video in different dimensions in a storyboard manner. A user can (1) select an image (or keyframe of a shot) as input. This image is further used by a feature-based search (2) that uses a feature-vector consisting of nine different features for comparison (in general color, texture, and transcript text). A user can manually tune the weighting of the different features. In the right part of the window, the results of the search are presented in a line-by-line and page-by-page manner (3) . The best result is presented at the left-top position of the first page and the worst result is presented at the right-bottom position of the last page. Furthermore, they use a relevance feedback technique in order to improve repeated search. On another tab (called \(NN^k\) network, (4)), the nearest neighbors of a selected image can be shown in a graph-like visualization. To provide temporal browsing they also use a fisheye visualization at the bottom of the window (5) in which the image-of-interest (selected on any view) is always shown in the center.

An extension of this approach is introduced by Ghoshal et al.\textsuperscript{76} Their interface, shown in Figure 14 is split into two main panels with the browsing panel taking up to 80% of the screen. The browsing tab (1) is divided into four tabs that provide different categories: Image & Feature Search, Content Viewer, Search Basket and \(NN^k\) keyframe browsing. In the Image & Feature Search tab (2), users can enter free text, named entities and visual concepts. Besides, they can specify the weighting of each textual and visual feature in using a sliding bar (3). The Content Viewer tab is divided into two tabs. On the left hand side (4), textual metadata of the last clicked keyframe is presented while on the right hand side, the full keyframe is shown. In the Search Basket tab, keyframes which are currently selected are displayed. The \(NN^k\) browsing tab shows these thirty keyframes which are nearest to the last clicked keyframe in the visual feature space.
Rautiainen et al. study content-based querying enriched with relevance feedback by introducing a content-based query tool. Their retrieval system supports three different querying facilities: query-by-textual-keyword, query-by-example and query-by-concept. The interface, shown in Figure 15, provides a list of semantic concepts a user can choose from. Textual based queries can be added in a text field on the top left hand side of the interface. Retrieved shots are represented as thumbnails of keyframes, together with the spoken text in the most dominant part of the interface. By selecting keyframes, users can browse the data collection using a cluster-based browsing interface. Figure 16 shows a screenshot of this interface. It is divided into two basic parts: On top is a panel displaying the selected thumbnail and other frames of the video in chronological order (1). The second part displays similar keyframes which have been retrieved by multiple content-based queries based on user-selected features (2). The keyframes are organized in parallel order as a similarity matrix, showing the most similar matches in the first column. This enables the user to browse through a timeline and see similar shots at the same time. Each transition in the timeline will automatically update the keyframes in the similarity matrix.

Figure 13. Video browsing/retrieval as proposed by Heesch et al.

Figure 14. Video browsing/retrieval as proposed by Ghoshal et al.
Campbell et al.\textsuperscript{79} introduce a web-based retrieval interface. Using this interface, users can start a retrieval based on visual features, textual queries or concepts. Figure 17a shows an example retrieval result. The interface provides functionalities to improve the visualization of retrieved keyframes by grouping them into clusters according to their metadata, such as video name or channel. Figure 17b shows an example grouping.

A similar approach is studied by Bailer et al.\textsuperscript{80} In their interface, shown in Figure 18, retrieval results are categorized into clusters. Single keyframes represent each cluster in the result list (1). Controls around the panel (2) depicting the search results allow the users to resize the presentation of these keyframes and to scroll through the list.

Foley et al.\textsuperscript{81} experimented in collaborative retrieval by introducing a multi-user system on a DiamondTouch\textsuperscript{82} tabletop device. Using the interface, a user can add keyframes as part of a search query and select which features of the keyframe shall be a reference for similar results. In their experiment, they asked sixteen novice users,
divided into eight pairs, to perform various search tasks. Each pair was sitting around the tabletop, facing each other. An additional monitor was used for video playback. Fig. 19 shows a screenshot of the interface. It provide facilities to enter a search query (1), to browse through keyframes (2), play a video shot (3), find similar keyframes (4), to mark keyframes as non-relevant (5) and to save the keyframes as a result (6).

Holthe and Ronningen\textsuperscript{83} presented a video browsing plug-in for a Web-browser, which can use hardware-accelerated graphics, if available, for improved browsing of video search results. Their browser allows compact views of preview images (in a 3D perspective) in order to increase the number of search results presentable on a single screen. When moving the mouse over a preview image the user can either zoom the image or start playback for the corresponding video segment, whereas the started video is presented in an overlay manner with the option of semi-transparent display.

Villa et al. presented the \textit{FacetBrowser},\textsuperscript{84} a Web-based tool which allows performing simultaneous search tasks within a video. A similar approach is introduced by Hopfgartner et al.\textsuperscript{85} The idea behind is to enable a user to explore the content of a video by individual and parallel (sub-)queries (and associated search results) in a way of exploratory search. A facet in that context is modeled as an individual search among others. The tool extracts speech transcripts from shots of the video for textual search. The results of a query are shown in a storyboard view where, in addition, a list of user-selected relevant shots for a particular query is shown as well. Moreover, the interface allows to add/remove search panels, to spatially move search panels, and to reuse search queries already performed in the history of a session.
Halvey et al.\cite{halvey} introduced ViGOR, a grouping oriented interface for search and retrieval in video libraries. The interface, shown in Figure 20 allows users to create semantic groups to help conceptualize and organize their results for complex video search tasks. The interface is split into two main panels. On the left hand side, users can enter a textual search query (1) and browse through the retrieval results (2). These results, represented by keyframes can be dragged and dropped to the example shots area (3) and will then be used as visual query. The right hand side of the interface consists of a workspace. In this workspace, users can create semantic groups (4), drag and drop keyframes in these groups (5) and retrieve visually similar shots by exploiting various low level visual features (6).

Adcock et al.\cite{adcock} presented an interactive video search system *MediaMagic* which has been used for TRECVID\cite{trecvid} over several years. The shot-based system allows searching at textual, visual, and semantic level. They use shot detection, color correlograms and SVM (Support Vector Machine) to analyze the content. A rich search interface is provided, which enables searching text queries, image queries and concept queries. In their interface they use *visual clues* to indicate which content item has been previously visited or explicitly excluded from search. Moreover, their system allows performing a multiple user collaborative search.
Neo et al.\textsuperscript{88} introduce an intuitive retrieval system called \textit{VisionGO} which is optimized for a very fast browsing of the video corpus. The retrieval can be triggered by entering a textual search query. Further, they can use keyboard shortcuts to quickly scroll through the retrieval results and/or to provide relevance feedback. The search query of later iterations is then further refined based on this feedback.

Most systems which have been introduced in this section support users in retrieving shots of a video. While this approach is useful in some cases, shots are not the ideal choice in other cases. Boreczky et al.\textsuperscript{89} argue for instance that television news consists of a collection of story units which represent the different events being relevant for the day of the broadcast. An example story unit from the broadcasting news domain is a report on yesterday’s football match, followed by another story unit about the weather forecast. Various systems have been introduced to provide users access to news stories (e.g., Lee et al.,\textsuperscript{90} Pickering et al.,\textsuperscript{91} Hopfgartner et al.\textsuperscript{92}). In all cases, stories are treated as a series of shots and the corresponding keyframes are visualized to represent a story. Figure 21 illustrates a representative interface as introduced by Hopfgartner and Jose.\textsuperscript{93} Users can (1) type in a search query and search results are ranked in either chronological order or based on their relevance to the search query.

Figure 21. Representative news video search interface (screenshot taken from live demo)

A summary of all introduced video retrieval interfaces is given in Table 3.
<table>
<thead>
<tr>
<th>Application</th>
<th>Browsing/Querying</th>
<th>Input Files</th>
<th>Struct. Unit</th>
<th>Video Dom.</th>
<th>Content Analysis</th>
<th>Visualization / Interaction</th>
</tr>
</thead>
<tbody>
<tr>
<td>Open Video Project$^{69}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>shot</td>
<td>all (c2)</td>
<td>shot boundary detection, text recognition</td>
<td>storyboard</td>
</tr>
<tr>
<td>Deng and Manjunath$^{70}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>shot</td>
<td>all (c2)</td>
<td>shot boundary detection, text recognition</td>
<td>storyboard</td>
</tr>
<tr>
<td>Komlodi et al.$^{61}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>shot</td>
<td>all (c2)</td>
<td>shot boundary detection, text recognition</td>
<td>storyboard</td>
</tr>
<tr>
<td>CueVideo$^{74}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>shot</td>
<td>news (c2)</td>
<td>shot boundary detection, text recognition</td>
<td>motion storyboard</td>
</tr>
<tr>
<td>Heesch et al.$^{74}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>shot</td>
<td>news</td>
<td>shot boundary detection, text recognition, visual retrieval</td>
<td>fisheye visualisation, storyboard</td>
</tr>
<tr>
<td>Rautiainen et al.$^{77}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>shot</td>
<td>news</td>
<td>shot boundary detection, text recognition, visual retrieval</td>
<td>storyboard</td>
</tr>
<tr>
<td>Campbell et al.$^{79}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>shot</td>
<td>news</td>
<td>shot boundary detection, text recognition, visual retrieval, concept filtering</td>
<td>storyboard, automatic grouping in clusters</td>
</tr>
<tr>
<td>Bailer et al.$^{80}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>shot</td>
<td>news</td>
<td>shot boundary detection, text recognition, visual retrieval, concept clustering</td>
<td>storyboard, automatic grouping in clusters</td>
</tr>
<tr>
<td>Foley et al.$^{81}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>shot</td>
<td>news</td>
<td>shot boundary detection, text recognition, visual retrieval</td>
<td>DiamondTouch</td>
</tr>
<tr>
<td>Villa et al.$^{84}$</td>
<td>yes/yes</td>
<td>no limited</td>
<td>shot</td>
<td>news</td>
<td>shot boundary detection, text recognition, visual retrieval</td>
<td>Facetted browsing</td>
</tr>
<tr>
<td>Halvey et al.$^{86}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>shot</td>
<td>news</td>
<td>shot boundary detection, text recognition, visual retrieval</td>
<td>storyboard, manual grouping</td>
</tr>
<tr>
<td>Adcock et al.$^{87}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>shot</td>
<td>news</td>
<td>shot boundary detection, color correlograms (with SVM)</td>
<td>storyboard, video player component, visual cues</td>
</tr>
<tr>
<td>VisionGo$^{88}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>shot</td>
<td>news</td>
<td>shot boundary detection, text recognition, visual retrieval</td>
<td>designed for quick access</td>
</tr>
<tr>
<td>Hopfgartner and Jose$^{93}$</td>
<td>yes/yes</td>
<td>no limit</td>
<td>story</td>
<td>news</td>
<td>story boundary detection, text recognition</td>
<td>storyboard, fisheye visualisation of story shots</td>
</tr>
</tbody>
</table>

Table 3. Overview of Video Retrieval Applications
4. VIDEO BROWSING APPLICATIONS BASED ON VIDEO SURROGATES AND UNCONVENTIONAL VISUALIZATION

Many papers can be found in the literature\textsuperscript{94–114} that describe video surrogates, which are alternative representations of the video content. The main purpose of video surrogates is to more quickly communicate the content of a video to the human observer. It is often used as a preview version for a video and should help the observer to decide whether the content is interesting or not. While such alternative representations are obviously important for video summarization many proposals have been made to use video surrogates also for video browsing and navigation.\textsuperscript{94} In this section we review applications using video surrogates for improved browsing or navigation.

The Mitsubishi Electric Research Laboratories (MERL) proposed several techniques for improved navigation within a video by novel content presentation. For example, the Squeeze layout and the Fisheye layout\textsuperscript{95} have been presented for improved fast-forward and rewind with personal digital video recorders (see Figure 22). Both layouts extract future and past DC images of the MPEG stream. In addition to the current frame the Squeeze layout shows two DC images at normal size, one taken 30 seconds from the future and another one taken 5 seconds from the past, and squeezes together the other frames in between. The Fisheye layout shows gradually scaled DC images (in the future and in the past) to the current frame, which is shown at normal size. Their evaluation has shown that subjects were significantly more accurate at fast-forward and rewind tasks with this display technique in comparison to a common VCR-like control set. In another paper of Wittenburg et al.\textsuperscript{96} the visualization technique has been generalized to Rapid Serial Visual Presentation (RSVP) (see Figure 23). Their model defines spatial layouts of keyframes in a 3D-like manner in different variations of trajectories. They evaluated the proposed RSVP technique for the purpose of video browsing by a user experiment with 15 subjects to compare it to the traditional VCR-like navigation set. The subjects were asked to answer questions like "Find the next commercials block." They showed that their approach can significantly outperform the VCR-like navigation set in accuracy. However, no significant difference was found in the task completion time. Shipman et al.\textsuperscript{97} described how the techniques of Wittenburg et al.\textsuperscript{96} have been adapted to a consumer product.

Campanella et al.\textsuperscript{98, 99} proposed a visualization of MPEG-7 low-level features (like dominant color, motion intensity, edge histogram, etc.) consisting of three parts (see Figure 24). The main part of the visualization consists of a Cartesian plane showing small squares representing shots, where each square is painted in the dominant color of the corresponding shot. The user can select a specific feature for both the x-axis and the y-axis, which immediately effects the positioning of those squares. For instance, motion intensity could be chosen for the y-axis whereas dominant color could be chosen for the x-axis (colors are ordered according to the hue value). That visualization scheme enables a user to detect clusters of shots and to determine the distances of such clusters, according to a particular feature. Below to the main window the shots are visualized in a temporal manner by painting stripes in the dominant color of each shot. Additionally, the right side shows keyframes of the currently selected shot. In a recent paper Campanella et al.\textsuperscript{115} describe an extended version of their tool with more interaction possibilities and additional views.

Axelrod et al.\textsuperscript{100} presented an interactive video browser that uses pose slices, which are instantaneous objects’ appearances, to visualize the activities within shots (see Figure 25). They perform a foreground/background segmentation in order to find the pose slices, which are finally rendered in a 3D perspective. Their video browser allows to simultaneously showing several positions of an object in a single video playback. Furthermore, the application enables a user to interactively control the viewing angle of the visualization.
Hauptmann et al. proposed the so called Extreme Video Retrieval (XVR) approach, which tries to exploit both maximal use of human perception skills and the systems’ ability to learn from human interaction. The basic idea behind is that a human can filter out the best results from a query and can tell the system which
results were right and which ones were wrong (a kind of *relevance feedback*). Therefore, they developed a *Rapid Serial Visual Presentation (RSVP)* approach, where keyframes of a query result are rapidly presented to the user who should mark correct results by pressing a key. By always presenting the keyframe at the same spatial location, their system avoids eye movements and, thus, minimizes the time for a user to perceive the content of an image. The frame rate, i.e. how fast the images are presented, is determined by the user. After the first run, a second correction phase with lower frequency is used to re-check marked keyframes. From this basic principle, extended versions have been implemented, where up to 4x4 images can be presented at the same time (*Stereo RSVP*), based on the natural parallelism of human binocular vision. In that case the user has 16 keys to mark a correct image in a grid-like presentation.

Eidenberger\textsuperscript{102} proposed a video browsing approach that uses similarity-based clustering. More precisely, a *Self-Organizing Map (SOM)*, which is a neural network that uses feed-forward learning is employed as a similarity-based clustering method. Visually similar segments of a video are grouped together and visualized in hierarchically organized index trees. He presented two types of index trees that can efficiently visualize the content of a video. While the *time index tree* shows the temporal structure of the video in a top-down manner, the *content index tree* shows the shot-based structure of the video in a bottom-up approach (i.e. the user starts browsing at a specific shot). The clusters are visualized as hexagonally shaped cells showing keyframes of shots. The user can interactively select a certain cell and step one layer deeper in the hierarchical tree structure to see more details of the selected shot. The number of layers in the tree depends on the length of the video. The user is able to switch between both different views at any time during the browsing process, which helps to preserve the browsing context. For the SOM-based clustering process several different types of MPEG-7 visual features, extracted for every frame, are used. Figure 26 shows SOM-based browsing with the content index tree that groups together similar shots into an iconic shot index.

A similar idea has been presented by Bärecke et al. who also used a growing SOM (Figure 27) to build a video browsing application called *VideoSOM*.\textsuperscript{103} Shots are non-temporally clustered according to a (probably color) histogram. Their tool provides a video player and several additional views at a glance:
(1) a self-organizing map window, showing keyframes of shot clusters gained through the learning phase,
(2) a list of shots (visualized by keyframes) according to a selected cluster,
(3) a timeline showing temporal positions of shots in the shot window.

Figure 27. Video browsing with VideoSOM

Goeau et al. proposed the so called Table of Video Contents (TOVC, see Figure 28) for browsing story-based video content like news, interviews, or sports summaries. Based on low-level features like color histograms in different color spaces, gradient orientation histogram, and motion model estimation (based on corner detection), they compute a similarity matrix which is further used for visualization. The 2D visualization uses a video backbone, either identified in a supervised way by an expert or automatically by finding clusters that contain the most covering frames. Every story is painted as a loop of keyframes originating from that backbone.

de Rooij et al. and Snoek et al. introduced the notion of video threads, where a thread is a sequence of feature-based similar shots from several videos in some specific order. They differentiate between

(1) visual threads having visual similarity,
(2) textual threads having similar textual annotations,
(3) semantic threads having semantically equivalent shots,
(4) time threads having temporal similarity,
(5) query result threads having similarity according to a query, and
(6) history threads consisting of shots the user has already visited.

Based on the abovementioned video threads they have implemented several different visualization schemes. The RotorBrowser (Figure 29) starts from an initial query result. The user can select a focal shot $S$ that is displayed (at bigger size) in the center of the screen. According to that focal shot $S$ the RotorBrowser provides several navigation paths by showing (parts of) all the video threads that contain $S$ in a star formation. As the RotorBrowser has been proven to be too overwhelming for non-expert users, the CrossBrowser (Figure 30) has been developed. The CrossBrowser does only provide horizontal and vertical navigation. For instance, the time
thread is visualized in the horizontal line while the visual similar shots of S are visualized in a vertical line. In the TRECVID 2006 evaluation of mean average precision, the CrossBrowser placed 2nd and the RotorBrowser placed 6th. The tool has been further extended in the ForkBrowser, which achieved even better results in the TRECVID 2008 evaluation.

Adams et al. published another interesting work called temporal semantic compression for video browsing. Their video browsing prototype (Figure 31) allows shot based navigation (bottom left in the figure), whereas only a few shots are shown at a glance containing the selected shot in the center. They compute a tempo function for every frame and every shot, based on camera motion (e.g. pan and tilt), audio energy, and shot length. The resulting function is plotted at the top right side in the figure. Their prototype enables a user to individually select a "compression rate" in order to shorten (i.e. summarize) the video. This function can be used by a simple slider or by directly clicking into the playback area, whereas the compression rate is derived from the vertical position and, in addition, the playback time position is selected by the horizontal position. Moreover, several different compression modes can be chosen. While linear compression simply speeds up playback, mid-shot...
constant takes a constant amount from the middle of a shot at constant playback rate. Pace proportional uses variable playback rate based on the frame-level tempo and interesting shots discard shots with low tempo values according to the selected compression rate.

Jansen et al. recently proposed to use VideoTrees (Figure 32) as alternatives to storyboards. A VideoTree is a hierarchical tree-like temporal presentation of a video through keyframes. The keyframes are placed adjacently to their parents and siblings such that no edge lines are required to show the affiliation of a node. With each depth level the level-of-detail increases as well (until shot granularity). For example, a user may navigate from a semantic root segment to one of the subjacent scenes, then to one of the subjacent shot groups, and finally to one of the subjacent shots. The current selected node in the tree is always centered, showing the context (i.e. a few of the adjacent nodes) in the surrounding area. In a user study with 15 participants they showed that the VideoTrees can outperform storyboards regarding the search time (1.14 times faster). However, the study also revealed that users found the classical storyboard much easier and clear.

Table 4 gives an overview of approaches reviewed in this section. All applications have been structured by the same criteria as used in Section 2.
Figure 32. Video browsing with the VideoTree.
<table>
<thead>
<tr>
<th>Browsing/Querying</th>
<th>Input Files</th>
<th>Struct. Unit</th>
<th>Video Dom.</th>
<th>Content Analysis</th>
<th>Visualization / Interaction</th>
</tr>
</thead>
<tbody>
<tr>
<td>MERL.95–97</td>
<td>yes/no</td>
<td>1</td>
<td>frame</td>
<td>all</td>
<td>DC-image extraction from MPEG files</td>
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<tr>
<td>Campanella et al. 99, 115</td>
<td>yes/no</td>
<td>1</td>
<td>shot</td>
<td>all</td>
<td>dominant color, motion, temporal position of shots</td>
</tr>
<tr>
<td>Axelrod et al. 100</td>
<td>yes/no</td>
<td>1</td>
<td>shot</td>
<td>all</td>
<td>foreground/background segmentation</td>
</tr>
<tr>
<td>Hauptmann et al. 101</td>
<td>yes/no</td>
<td>1</td>
<td>shot</td>
<td>all</td>
<td>shot boundary detection</td>
</tr>
<tr>
<td>Eidenberger 102</td>
<td>yes/no</td>
<td>1</td>
<td>frame</td>
<td>all</td>
<td>shot boundary detection, MPEG-7 visual feature extraction (and similarity-based clustering with SOMs)</td>
</tr>
<tr>
<td>Bärecke et al. 103 (VideoSOM)</td>
<td>yes/no</td>
<td>1</td>
<td>shot</td>
<td>all</td>
<td>shot boundary detection, clustering based on (color) histograms</td>
</tr>
<tr>
<td>Goeau et al. 104 (TOVC)</td>
<td>yes/no</td>
<td>1</td>
<td>story</td>
<td>news</td>
<td>color histograms, gradient orientation histograms, motion model estimation</td>
</tr>
<tr>
<td>de Rooij et al. 105 and Snoek et al. 106 (RotorBrowser, CrossBrowser)</td>
<td>yes/no</td>
<td>N</td>
<td>shot</td>
<td>all</td>
<td>shot boundary detection, visual cues, concepts</td>
</tr>
<tr>
<td>Adams et al. 107</td>
<td>yes/no</td>
<td>1</td>
<td>shot</td>
<td>all</td>
<td>shot boundary detection, analysis of camera motion and audio energy</td>
</tr>
<tr>
<td>Jansen et al. 108</td>
<td>yes/no</td>
<td>1</td>
<td>shot</td>
<td>all</td>
<td>shot boundary detection,</td>
</tr>
</tbody>
</table>

Table 4. Overview of Video Browsing Applications using Video Surrogates
5. CONCLUDING REMARKS

We have reviewed video browsing approaches that have been published in the literature within the last decade. We classified existing approaches into three different types:

- interactively browsing and navigating through the content of a video in a video player like style,
- browsing the results of a video retrieval query (or a large video collection), and
- video browsing based on video surrogates.

Our review has shown that research in video browsing is very active and diverse. While a few approaches simply try to speed up the playback process for a video, several others try to improve the typical interaction model of a video player. In fact, the navigation features of a standard video player are still very similar to those of analogue video cassette recorders invented in the 1960s (apart from faster random access). Even popular online video platforms use such primitive navigation functions. The main reason is surely that most users are familiar with the usage of simple video players. Section 2 has revealed that many other ways are available to improve common video players yet keeping interaction simple. Many other approaches try to optimize the visual presentation of a large video collection or a number of search results. At one hand the storyboard has been established here as a standard means to display a large number of keyframes and it is used in most video retrieval applications. On the other hand, Section 4 has shown that video surrogates can more effectively convey video content information. Appropriate video surrogates can significantly improve the performance of video browsing and video retrieval applications. The reason is that human users can easily and quickly conclude content correlations from appropriate visualizations and use their personal knowledge and experience to improve the search process. Nevertheless, to design generally usable video surrogates might be hard. It is obvious that video surrogates need to be specifically designed for several different types of video content and a lot of research need to be performed into that direction.

This review has not only shown that the user interfaces of video browsing applications are very diverse, but also the methods used for content analysis are very different. While some methods use no content analysis at all, which has the non-negligible advantage of a short "start-up delay" for a new video from the user perspective, others perform intensive multi-model analysis. In general, we can conclude that the content analysis technique to be used is highly dependent on the video domain. For news videos most approaches use text recognition and a few apply face detection. In contrast for sports motion and speech analysis is typically used. If the application should be useable in several domains color and motion features are often employed. The video domain also determines content segmentation. While shots are typically used for general purpose applications, story units are the structuring element for news domain applications.

Future challenges are to further assist users in video browsing and exploration. Intelligent user interfaces are required that do not only visualize the video content but also adapt to the user. In an video retrieval scenario, this adaptation can be achieved by employing relevance feedback techniques. Moreover, considering the increasing amount of diverse user generated video content, e.g. on social networking platforms, another challenge is how interfaces can deal with this low-quality material.
REFERENCES


