A meeting recorder captures multimodal information of a meeting. Subsequent analysis of the information produces scores indicative of visually and aurally significant events that can help identify significant segments of the meeting recording. Textual analysis can enhance searching for significant meeting segments and otherwise enhance the presentation of the meeting segments.

17 Claims, 14 Drawing Sheets
OTHER PUBLICATIONS


* cited by examiner
FIG. 1
Fig. 4
BERNA: But, um, let me see if I can get this document. Is this a GW copier?
DEREK: Yeah, this is.
DAR-SHYANG: Actually it is. How are you gonna get your document?
BERNA: Oh, I--I'm going to use the ...

BERNA: Let's talk about it a little bit more.
DEREK: What do you--what do you think the cultural factors are, Dar?
DAR-SHYANG: Um, I don't know.
DEREK: I think color suggests, you know, individuality and, uh-- DAR-SHYANG: Right.
DEREK: --certainly allows you to stand out a little bit. DAR-SHYANG: Right.
BERNA: It is also the pure fact that we did our study about two years ago--
DEREK: Hmm. BERNA: --and, uh, you guys conduct your study recently, right?
FIG. 6

FIG. 7
Fig. 8B

<table>
<thead>
<tr>
<th>Start_T</th>
<th>End_T</th>
<th>Azimuth</th>
<th>Elevation</th>
<th>Zoom</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.00</td>
<td>4.85</td>
<td>193</td>
<td>30</td>
<td>1.0</td>
</tr>
<tr>
<td>4.86</td>
<td>10.57</td>
<td>320</td>
<td>30</td>
<td>1.0</td>
</tr>
<tr>
<td>10.58</td>
<td>15.36</td>
<td>111</td>
<td>30</td>
<td>1.0</td>
</tr>
<tr>
<td>15.37</td>
<td>20.22</td>
<td>320</td>
<td>30</td>
<td>1.0</td>
</tr>
<tr>
<td>20.23</td>
<td>30.00</td>
<td>111</td>
<td>30</td>
<td>1.0</td>
</tr>
</tbody>
</table>

.....

Sound localization

Extract several snapshots of the speaker

Face localization

Find the best shot

Fig. 9
Meeting Description:
Multimedia dynamic interactive group meeting
regular MDA meeting
Date: 10/30/07
Time: 10:00 AM to 12:31
Location: After meeting room

Attendees:
[Images of attendees]

Audio to keyframes, 1018
Transcription, 1014

Fig. 10
MULTIMODAL ACCESS OF MEETING RECORDINGS

CROSS-REFERENCES TO RELATED APPLICATIONS

This application is related to the following commonly owned and co-pending U.S. patent applications which are hereby incorporated by reference for all purposes:

U.S. patent application Ser. No. 60/52,1252 titled “Method and System for Information Management to Facilitate the Exchange of Ideas During a Collaborative Effort,” filed Mar. 8, 2000;

U.S. patent application Ser. No. 10/081,129 titled “Multimedia Visualization & Integration Environment,” filed Feb. 21, 2002; and


BACKGROUND OF THE INVENTION

The present invention relates generally to multimedia meeting recordings and more specifically to access of multimedia meeting recordings.

Progress in the business world typically finds its beginnings with a series of meetings. Meetings are usually drawn out unstructured affairs, consisting of mostly irrelevant information. However, there are usually one or two defining moments that occur during a meeting which can propel the enterprise in the forward direction toward success, or if missed, can result in yet another failed business venture.

Many businesspersons visit remote locations and participate in meetings with different people on a regular basis. A common task that must be performed at some time after the meeting is the creation of a summary of what happened during the meeting. The summary may include reports of who said what, the ideas that were conceived, the events that occurred, and the conclusions that were reached. Often times, it is not just the specific conclusions but also the reasons they were reached and the points of view expressed by the meeting participants that are important.

Producing an accurate meeting summary is a time-consuming and error-prone process, especially if the only record available is one’s own memory, perhaps supplemented with hastily handwritten notes. A commonly used portable memory aid is the audiocassette recorder. It can be effective, but lacks the ability to capture important events that could be helpful later, such as gestures, images of participants, body language, drawings, and so on. An easy-to-use method for incorporating video data would help solve this problem.

Meeting recordings can help. Capturing the content of meetings is useful in many respects. Recordings of a meeting can capture the meeting activity and then later be reviewed as needed to refresh one’s memory. Knowing that a meeting is being recorded allows the participants to more effectively concentrate on the meeting discussion since the details can be later reviewed in an offline manner. Audiovisual recordings of meetings provide the capability for reviewing and sharing meetings, clarifying miscommunications, and thus increase efficiency.


However, people generally prefer not want to watch a recorded meeting from beginning to end. Like the meeting from which the recording was produced, recorded meetings are not amenable to a hit-or-miss search strategy. After fast-forwarding a few times in a meeting while looking for something, most people will give up unless what they are seeking is important enough to suffer the tedium.

More often than not, people are only interested in an overview of the meeting or just the interesting parts. Enabling efficient access to captured meeting recordings is essential in order to benefit from this content. Searching and browsing audiovisual information can be a time consuming task. Two common approaches for overcoming this problem include key-frame based representations and summarization using video skims. Key-frame based representations have proven to be very useful for video browsing as they give a quick overview to the multimedia content. A key-frame based technique is described by S. Uchihashi, J. Foote, A. Girkensohn, and J. Boreczky, Video Manga: Generating Semantically Meaningful Video Summaries, ACM Multimedia, (Orlando, Fla.) ACM Press, pp. 383-392, 1999.

On the other hand, video skims are content-rich summaries that contain both audio and video. Efficiently constructed video skims can be used like movie trailers to communicate the essential content of a video sequence. For example, A. Waibel, M. Bett, et al., Advances in Automatic Meeting Record Creation and Access, Proceedings of ICASSP, 2001 propose summarization of meeting content using video skims with a user-determined length. The skims are generated based on relevance ranking and topic segmentation using speech transcripts.

A summarization technique for educational videos based on shot boundary detection, followed by word frequency analysis of speech transcripts, is suggested by C. Taskiran, A. Amir, D. Ponceleon, and E. J. Delp, Automated Video Summarization Using Speech Transcripts, SPIE Conf. on St and Ret. for Media Databases, pp. 371-382, 2002.

A method for summarizing audio-video presentations using slide transitions and/or pitch activity is presented by He, L., Sanocki, E., Gupta, A., and Grudin, J., Auto-summation of audio-video presentations, In Proc. ACM Multimedia, 1999. The authors suggest a method of producing presentation summaries using video channel, audio channel, speaker’s time spent on a slide, and end user’s actions. The authors discuss the use of pitch information from the audio channel, but their studies showed the technique as being not very useful for summary purposes.

Instead, they indicate that the timing of slide transitions can be used to produce the most useful summaries.

Motion content in video can be used for efficiently searching and browsing particular events in a video sequence. This is described, for example, by Pingali, G. S., Opalach, A., Carlbom, I., Multimedia Retrieval Through Spatio-temporal Activity Maps, ACM Multimedia, pp. 129-136, 2001 and by Divakaran, A., Vetro, A., Asai, K., Nishikawa, H., Video Browsing System Based on Compressed Domain Feature Extraction, IEEE Transactions on Consumer Electronics, vol. 46, pp. 637-644, 2000.

As a part of the Informedia™ project, Christel, M., Smith, M., Taylor, C. R., and Winkler, D. Evolving Video Skims Into...
SUMMARY OF THE INVENTION

A method for creating video skims is based on audio and visual activity analysis together with text analysis. In an illustrative embodiment of the invention, detection of important visual events in a meeting recording can be achieved by analyzing the localized luminance variations in consideration with the omni-directional property of the video captured by our meeting recording system. In another illustrative embodiment, audio activity analysis is performed by analyzing sound directions—indicating different speakers—and audio amplitude. A further aspect of the present invention is incorporation of text analysis based on the Term Frequency—Inverse Document Frequency measure. The resulting video skims can capture important segments more effectively as compared to the skims obtained by uniform sampling. It can be appreciated that the techniques according to the present invention can be applied to any multimedia recording, wherein meeting recordings constitute only a subset of the broader category of multimedia recordings.

DESCRIPTION OF THE SPECIFIC EMBODIMENTS

From the following it will be appreciated that the present invention has broad application generally to multimedia information. However, to facilitate a discussion of the invention sufficient to enable its practice, the present invention will be described in the context of a meeting recording system.
A meeting recording system can be an omnidirectional camera and microphone arrays. Subsequent processing of the data by known sound localization techniques, either in real-time or off-line, can be employed to find the direction of the sound source. In a particular embodiment of the invention, the video recording system can also provide sound localization information. For example, sound data can be collected with two cones of unequal thickness, if they intersect at all. The meeting recorder can comprise a video recording system to produce a visual recording of the meeting participants and possibly any other visual activity that may be present such as a video presentation made during the meeting. In one embodiment of the invention, for example, the video recording system can be an omnidirectional camera having a parabolic mirror for capturing a panoramic view of the meeting room. Such cameras are known. The video stream comprises doughnut-like images.

In a particular embodiment of the invention, the audio recording system can also provide sound localization information. For example, sound data can be collected with microphone arrays. Subsequent processing of the data by known sound localization techniques, either in real-time or off-line, can be employed to find the direction of the sound. In a particular embodiment of the invention, sound localization is performed in real-time to avoid the need of handling and saving multiple channels of audio data. This approach may be suitable in a compact configuration such as in a portable system, for example.

The audio signal is processed in segments of 25 msec. Since we are interested only in human speech, segments that do not contain speech in at least one of the sound channels can be ignored. Following speech detection, 360-degree sound localization can be determined in the following manner. For each pair of microphones that are diagonally situated, an angle between 0 and 180 degrees is calculated based on phase difference. This angle defines a cone of confusion centered at the midpoint of the diagonal. In theory, the intersection of two cones computed from both diagonal pairs defines the azimuth and elevation of the sound source. Unfortunately, the angle computed by each pair is not perfect. Moreover, phase difference measured on a finite sampling rate over a small baseline is discrete, and the angular resolution over all directions is non-uniform. Higher resolution is obtained near the center, and is lower towards both ends. Therefore, we need to compute the intersection of two cones of unequal thickness, if they intersect at all. Furthermore, we want to take into consideration the confidence associated with each angle estimate.

To resolve these issues, we use an accumulator over the parameter space of azimuth by elevation. Azimuth varies from 0 to 360 degrees and elevation varies from 0 to 90 degrees. For each possible (azimuth, elevation) pair covered by each cone, its entry is incremented by the confidence associated with the cone. The highest scoring entry in the accumulator corresponds to the best parameter estimate. All entries in the accumulator are decayed by a factor at the end of each segment. However, in trying to estimate both azimuth and elevation, we found the solution unstable and sensitive to the quantization chosen. Furthermore, it does not account for the fact that sound sources close to the middle are detected more accurately than those close to either end. Therefore, the scores at all elevations are summed up for each azimuth, and the best azimuth is returned if its score exceeds a threshold. Then, for each segment where speech is found, a triplet of time-stamp, angle, and score, denoted by (t, θ, w), is written to a file. We observed this process is capable of performing in real-time, consuming approximately 25% to 40% CPU load on a 933 MHz PC.

Text capture devices can be incorporated into the meeting recorder. For example, a scanner can be provided to allow scanning of documents that are handed out during the meeting. Notes generated during the meeting can be scanned in. The text capture can include associating some form of information identifying the originator of the text being captured. As a catch-all, it can be appreciated that other capture devices can be incorporated into the meeting recorder. For example, slides from a slide presentation can be captured. Where a whiteboard is used, one can appreciate that the contents of the whiteboard can be captured using appropriate known whiteboard capture techniques. Meeting participants can use graphics tablets or similar input devices to record handwritten notes. Such devices can facilitate the capture of notes by obviating the step of scanning in notes written on conventional media such as paper.

FIG. 1 shows the meeting recording and access system exemplar shown in FIG. 1 comprise a meeting recorder component for capturing a variety of media information that can be produced during the activity of a meeting. For example, the meeting recorder can comprise a video recording system to produce a visual recording of the meeting participants and possibly any other visual activity that may be present such as a video presentation made during the meeting.

The video recording component can be part of the meeting recorder. Additional processing can be performed to generate sound localization information, which can occur off-line using a separate computer system. In a given implementation, the audio processor and the video processor can be the same processor.

FIG. 1 shows that the information captured by the capture components collectively represented by the data line can be stored in some sort of data store represented functionally as a storage component for subsequent processing and retrieval. It is noted that this component is only a logical representation of the actual storage hardware, as are the other components shown in the figure.
storage devices may be used to store different data, including captured data and processed data.

A metrics generator functionality 106 produces a variety of metrics from the various information collected in accordance with the invention (additional details about the computations provided below) and then can be stored in the storage component 112. In one embodiment of the invention, the information can feed directly to the metrics generator from the various data capture device 104a-104d. In another embodiment, the source data for the metrics generator can be obtained from the storage component 112 in an off-line manner.

An access engine functionality 124 provides access to the information contained in the storage component. As will be explained below, the captured information can be selected in accordance with the invention based on various combinations of the computed metrics. An interface engine functionality 122 provides an appropriate user interface to the access engine functionality. For example, a graphical user interface (GUI) may be appropriate to facilitate browsing the recorded meeting information. It is noted here that a “user” need not be a conventional human user. Another machine can serve as the “user” of the present invention. In such a case, the interface engine functionality can be defined by an appropriate data exchange protocol. The interface engine functionality can simply be a suitable application programming interface (API), defining a library of utilities to facilitate implementing a suitable GUI, or a machine interface, or even a simple command line interface (CLI).

The discussion will now turn to various metrics that can be produced by the metrics generator functionality 106 according to the present invention. Consider first the video information collected by the video capture functionality 104a. In accordance with an embodiment of the present invention, the video is captured with an omni-directional camera and converted to the MPEG-2 data format for storage and subsequent access.

In a video of a typical meeting sequence, most of the time there is minimal motion. High motion segments of video usually correspond to significant events. For example, a participant getting up to make a presentation, someone joining or leaving the meeting, or just making gestures, could be important for searching and recalling some segments of the meeting.

In accordance with the invention, a visual significance measure is generated based on local luminance changes in a video sequence. A large luminance difference between two consecutive frames is generally an indication of a significant content change, such as a person getting up and moving around. However, insignificant events, such as dimming the lights or all the participants moving slightly, may also result in a large luminance difference between two frames. In order to reduce the likelihood of identifying such events as being significant, the visual significance measure, according to an embodiment of the invention, can be determined by considering luminance changes occurring in small windows of a video frame rather than a single luminance change of the whole frame.

The luminance changes can be computed by computing the luminance difference between the consecutive intra coded (I) frames in the MPEG-2 compressed domain. We employ I-frames because the luminance values in I-frames are coded without prediction from the other frames, and they are therefore independently decodable. We compute luminance differences on the average values of 8x8 pixel blocks obtained from the DC coefficients. The DC coefficients are extracted from the MPEG bit stream without full decompression. Average values of the 8x8 pixel blocks are found by compensating for the DC prediction and scaling. In a particular implementation, the DC coefficients are obtained from the bit stream using a modified version of the publicly available TMN MPEG-2 decoder available at the FTP site: ftp://tmn-ip.cs.berkeley.edu/pub/multimedia/mpeg2/soft-ware/).

We modified the decoder to only parse the intra coded frames and to skip some of the decoding operations (e.g., Inverse DCT, Inverse quantization, motion compensation). Because full decoding and reconstruction is not performed, a one hour video can be processed in less than 4 minutes on a 1700 MHz computer. The effects of the operation can be observed with reference to the frames shown in FIGS. 2A and 2B. FIG. 2A shows an original frame of MPEG-2 video. FIG. 2B illustrates a reconstruction of the frame using only the DC coefficients.

Since the video is recorded using an omni-directional camera, the image is doughnut-shaped. As can be seen in FIG. 3, for example, the pixels (or the DC coefficients) in the outer parts of the doughnut video contain less object information (i.e. more pixels per object), therefore their weight is less than the pixels (or the DC coefficients) that are towards the inner parts. Thus, when computing the frame differences, the pixel values (or the DC coefficients) are weighted according to their location to compensate for this. It can be appreciated of course that for a rectangular-shaped video frame, the weight values w(r) are unity.

The weights of the pixels are assigned to compensate for this ratio when computing the frame pixel differences. The assignment of weights is accomplished by considering the parabolic properties of the mirror as follows:

\[ w(r) = \frac{1}{\cos^{-1}\left[ \frac{1}{1 + \frac{4r^2}{R_{max}^2}} \right]} \]

where \( r \) is the radius of the DC coefficient location in frame centered polar coordinates and \( R_{max} \) is the maximum radius of the donut image. The coefficients that do not contain any information (the location that corresponds to outside of the mirror area) are weighed zero.

We employ a window size of 9x9 DC coefficients, which corresponds to a 72x72 pixel area, though it can be appreciated that other suitable window sizes corresponding to different pixel areas can be used. The weighted luminance difference is computed for every possible location of this window in a video frame. A local visual activity measure (score), \( V_{av} \), can be defined as the maximum of these differences as follows:

\[ V_{av} = \max \left[ \sum_{x=-L/2}^{L/2} \sum_{y=-H/2}^{H/2} \left( \sqrt{(x-n)^2 + (y-m)^2} \right) w_{av}(n,m) \right] \]

where \( W_{av} \) is the weighted local activity frame (in number of pixels).

The inverse of the DC block at location \( r \) (in polar coordinates); and
video is encoded using MPEG encoding, so the video frame turn can serve as a basis for identifying visually significant events. As with visual events can be achieved if we compensate for the information from our meeting recorder system.

...the corresponding computed visual activity score plotted at point 402a exhibits a large value, illustrating the correlation between the computed metric and the visual event. Similarly, the video frames showing a person leaving the meeting room at event (c) also corresponds to a large value at point 402c, again indicative of a visually significant event. Events (d) and (e), showing a person entering the meeting room at event (c) also correspond to a large value at point 402e, again indicative of a visually significant event. Events (f) and (g), showing the large moving object due to the change in perspective. Exclusion of such segments from the category of important events, which can be comprised of arguments and disagreements among meeting participants. Though high amplitude audio resulting from a raised voice provides a good indication of the presence of emotion, it was discovered that amplitude by itself is not a sufficient indication of a significant audio event. For example, the sound of a door slamming or the noise from a heavy object falling to the floor do not constitute significant audio events.

...it is noted that Eqn. 2 is equally applicable to non-MPEG encoded video. For example, unencoded video, it can be appreciated that the foregoing operations can be performed on pixels rather than DC blocks. More generally, depending on the specific encoding, a “unit pixel block” can be defined as being the smallest image unit on which Eqn. 2 operates. Thus, in the case of MPEG video, the “unit pixel block” is a DC block of 8x8 pixels. For unencoded video, however, the “unit pixel block” can be a single pixel. A unit pixel block can be square (e.g., 2x2 pixel block, or more generally an N x N block), it can be rectangular (e.g., M x N, M > N), or some other shape.

FIG. 4 shows a plot 402 of computed local visual activity scores versus time for a meeting video. As shown in the figure, many of the peaks of the visual activity score correspond to significant visual events. For example, at event (a), the video frames taken at time t and (t+Δt) show a person taking his place at the table, which can be deemed to be visually significant event. The corresponding computed visual activity score plotted at point 402a exhibits a large value, illustrating the correlation between the computed metric and the visual event. Similarly, the video frames showing a person leaving the meeting room at event (c) also corresponds to a large value at point 402c, again indicative of a visually significant event. Events (d) and (e), showing a person entering the room at event (c) also correspond to a large value at point 402e, again indicative of a visually significant event. Events (f) and (g), showing a person taking his place at the table, which can be deemed to be visually significant event. Thus, in the case of MPEG video, the unit pixel block can be square (e.g., 2x2 pixel block, or more generally an N x N block), it can be rectangular (e.g., M x N, M > N), or some other shape.

...This metric is a measure of the conversational activity among the participants in the meeting. G(n) can be any suitable smoothing filter. For example, in an embodiment of the invention a Gaussian filter is employed. C(t) is the number of changes in the sound direction or the number of speaker changes, measured in the time frame of [t-1/2, t+1/2]. The time index t can be any time unit; here we use seconds. For example, if the sound direction was constant between [t-1/2, t+1/2] seconds, C(t) takes value “0”. If there was one sound direction or speaker change, then C(t) takes value “1” and so on.

Audio activity measure, \( U_a(t) \), is defined as:

\[
U_a(t) = \alpha \sum_{k=-W/2}^{W/2} |X(n)|^2 + k,
\]

where \( S_a(n) \) is speaker activity at time index t, \( f_s \) is the audio sampling frequency, and \( X(n) \) is the audio amplitude of the n\(^{th} \) sample. In accordance with this particular embodiment of the invention, the audio activity measure is the basis for the audio significance score.

FIG. 5 illustrates experimental results of recordings processed in accordance with the invention; in this particular example, audio clips from a staff meeting recording. The figure includes a sound localization plot 502, showing locations of sound origination recorded over time. Here, rather than the absolute values of the sound directions, the changes in sound directions are significant. C(t), which is used to compute \( S_a(t) \), is computed based on a count of changes in the sound direction in a \([t-1/2, t+1/2]\) window. The sound localization plot graphically illustrates the changes in sound direction which occur during a conversation among many people. Audio activity is represented by an audio activity plot 504 which is a plot of audio activity metrics \( U_a(t) \) versus time t. In the example shown, a transcript of each of the audio clips from the meeting discussion is provided in the dialog boxes 512, 514.

The audio activity plot 504 exhibits high regions 522 and 524, which correlate with the speaker activity shown in the dialog boxes. For example, the dialog box 512 indicates an exchange of ideas among four speakers. The speaker activity is illustrated in the region 522 of the sound localization plot 502, indicating a local increase in the rate of change of the sound location. The speaker activity is shown by the peaks in the corresponding region of the audio activity plot. Similarly, the dialog box 514 indicates an even more animated interchange of ideas. This can be seen by inspecting...
the corresponding period of time on the sound localization plot 502 at region 524. The corresponding region of time on the audio activity plot 504 shows a region of peaks in the audio activity metrics.

It can be appreciated therefore that by analyzing various parameters of these audio plots, significant audio events can be identified. For example, when an audio localization plot is produced from initial experiments with several recordings of staff meetings, presentations, and brainstorming sessions, the peaks in the audio activity plot 504 bear a correspondence to the audio clips with a high degree of meeting participant interactions and few silent periods.

Still another aspect of the metric generation functionality 106 shown in FIG. 1 is generation of textual scores. In accordance with the invention, the audio recording of a meeting can be transcribed to produce a rich source of text. Language analysis techniques are commonly used to summarize documents and audio transcriptions. Here, we compute the well-known Term Frequency-Inverse Document Frequency (TF-IDF) on meeting transcriptions in order to find segments that contain important keywords. TF-IDF is defined as TF-IDF = df where tf is the frequency of a word in a document and df is the frequency of the same word in a collection of documents. For a transcript of a meeting recording, the collection of documents comprises the collection of transcripts from various meetings.

This measure is employed in our system as follows, referring to FIG. 7. A transcript 704 is produced from the audio component 702 of the meeting recording. A TF-IDF score is produced for each word in a set of words 706 taken from the transcript. The words having the highest TF-IDF scores define a set of keywords 708. For example, the keywords can be those words 706 having TF-IDF scores which exceed some pre-selected threshold value. It can be appreciated that other techniques can be used to define the set of keyword; however, TF-IDF is recognized as a technique for providing reliable performance in many applications. In order to find when a given keyword most frequently occurs, we divide the audio transcriptions into audio segments 710 of some duration (e.g., transcriptions of 10-second audio clips) and compute a document occurrence score DOk as follows:

\[ DOk(i) = \sum_{n=1}^{W} G(n)O_k(i+n), \]

where i is the audio segment number;

G(n) is the smoothing filter coefficient;

W is the length of the smoothing filter; and

O_k(i) is the number of occurrences of the keyword k in the audio segment.

The audio segment with the highest DOk value is defined as the keyword audio segment 712 for keyword k. These mapped audio segments 712 can be used to enhance accessed segments (e.g., video skims) of the meeting recording. For example, video clips corresponding to the keyword audio segments can be added to the video skim.

FIG. 6 is a high level generalized flow diagram, illustrating the basic processing for identifying and accessing summaries of a meeting recording comprising significant segments from the meeting recording in accordance with the invention. Our goal is to find meeting segments that communicate the salient content of a meeting most efficiently. The process begins with a step 602 of obtaining a recording of the meeting. As can be appreciated from the foregoing, a meeting recording can comprise various modalities of recorded information, including video, audio, and textual information. In accordance with a particular implementation of the invention, the recorded meeting information is reduced to digital form for subsequent processing. For example, video information can be encoded in a suitable MPEG format, audio can similarly be digitized, and textual information can be converted by appropriate optical character recognition (OCR) software.

In a step 604, significance scores are computed from the captured recorded meeting information. For example, in an embodiment of the invention, a visual significance score \( V_a \) can be produced for each video frame of the video component of the captured recording meeting based on local luminance metrics computed for that frame, such as discussed above. Similarly, an audio significance score \( U_a \) can be computed in the manner described above, using sound localization information contained in the audio component of the recording meeting. Transcripts can be generated from the audio component and processed so that each of a number of keywords identified in the transcript can be associated with one or more audio clips (keyword audio segments) that are deemed to be significant with respect to that keyword.

Next, in a step 606, the computed scores generated in step 604 can be ranked, e.g., sorted by magnitude. In accordance with an embodiment of the present invention, the visual significance scores \( V_a \) and audio significance scores \( U_a \) are sorted. Similarly, computed keyword audio segments (those with the highest DOk values) are sorted according to the TF-IDF of the keywords.

Access to the meeting recording can be provided in the form of extracted portions of the meeting recording, in a step 608, based on one or more of these scores. In a particular implementation of the invention, the extracted content can be presented as a video skim (comprising, for example, video and audio) that is played on a suitable display; e.g., an audio-video capable system such as a PC with speakers. A video skim can be produced based on a visual significance score. Alternatively, the video skim can be based solely on an audio significance score. Still another alternative is a video skim based on a combination of two or more of these and other scores, or on a composite score computed as a function of two or more of these and other scores.

Suppose the recorded meeting is accessed as a one minute video skim based solely on a given visual activity score. A video frame having that score is selected from the recorded meeting data and can be the basis from which the video skim is produced. In this case, the video skim might comprise a one minute segment of consecutive video frames (video clip) from the recorded meeting starting with the selected video frame. Alternatively, the video clip might be selected from the recorded meeting such that it ends with the selected video frame. A segment of the recorded meeting can be selected which includes the selected video frame somewhere in the video clip. If appropriate, the video clip might even be a segment of the recorded meeting that precedes (or follows) the selected video frame by some amount of time. It might be desirable to compose the one minute video skim by combining a series of shorter video clips of equal or unequal duration. The video clip can be presented as video-only, absent a corresponding audio track. Such a presentation
might be desirable, depending on the purpose of the users. Of course, the video skim can include the corresponding audio track.

Depending on performance, storage capacity, and other such criteria, these video skims can be produced on a demand basis; that is, only when a visual significance score is presented. On the other hand, a video skim for some or all of the visual significance scores can be generated offline and stored for later retrieval, thus allowing faster access albeit possibly at the expense of providing substantial storage capacity.

Content of the recorded meeting can be accessed (e.g., a one minute video skim presentation) based solely on a given audio significance score. Again, suppose a one minute video skim is desired. The time index corresponding to the given audio score is determined. In this case, the video skim can comprise a one minute segment of the video (video clip) taken from the recorded meeting based on that time index, including possibly the audio track. The video clip can begin or end at that time index, or the video clip can span a period of time that includes the time index. The video clip might even be for a period of time that is earlier than the time index or is later than the time index. It may be desirable to compose the one minute video skim by combining a series of shorter length video clips of equal or unequal duration. Also, it is possible to store video skims for some or all of the audio significance scores in an offline manner as discussed above.

It can be appreciated that as an alternative to composing and presenting a video skim, an audio-only segment corresponding to the given audio significance score can be presented; though such a presentation may not be as effective as a presentation that includes the video track. However, in situations where no corresponding visual information is available, an audio-track only presentation would be appropriate.

The recorded meeting can be accessed by presenting a video skim that is generated based on a given visual significance score and a given audio significance score. For example, consider again that a one minute video skim is desired. The video skim can comprise a first segment of video based on the visual activity score as discussed above and a second segment of video based on the audio score as discussed above. The segments can be of equal or unequal duration.

It can be appreciated that where two non-consecutive video clips are played back, it may be desirable to “stitch” together the clips in order to smooth out the transition between clips to minimize the distraction of a sudden scene change. For example, a wipe transition (e.g., right to left) can be applied to the video tracks to transition from one clip to the next. Of course, other transitions can be used. Transitioning of the corresponding audio tracks from one video clip to the next can be achieved by audio fades or the like. If any two clips are close in time or if they overlap, they can be merged into a single clip to achieve continuity.

Continuing with FIG. 6, consider an example where a user might request the most significant video skim of the meeting (step 610) of say five minutes in duration. The video skim can be produced by taking the highest valued visual significance score from the ranked list of scores (step 604) and accessing the corresponding video frame. The video skim can then be produced from that video frame in the manner described above. In the case where two or more scores have the same highest value, a random selection can be made to select from among the two or more corresponding video frames as the basis for generating the video skim. Another selection criterion might be to choose the earliest video frame, or the latest video frame, or the middle-most video frame. A request can be presented to the user to make the selection of video frame. Still another resolution might be to use all the video frames combined in the manner discussed below to produce the video skim.

Further in accordance with the invention, the video skim can be produced from plural video frames. In this approach, the first N highest-valued visual significance scores can be selected from the ranked list of visual activity scores (step 604). A video clip can be produced for each of the selected N video scores in the manner described above. The time duration of each video clip can be equal or unequal. The video clips can then be stitched together using suitable transitioning techniques to produce a video skim of the desired duration; e.g., five minutes. The value of N can be a user-specified quantity. It can be appreciated that the duration of each video clip comprising the video skim are dependent on the duration of the video skim and the value of N. Generally, the following relation should hold true:

\[ D \approx \sum_{i=0}^{i=N-1} d_i \]

where \( D \) is the duration of the requested video skim, \( N \) is the number of video clips comprising the video skim, and \( d_i \) is the duration of the \( i \)-th video clip.

The approximation sign (\(~\)\) reflects the possibility that the total duration of the final video skim may be longer than (or shorter than) the duration \( D \) of the desired video skim, owing largely to the application of transition effects between discontinuous segments. However, due to merging of close-in-time segments, and other such factors, the total time might come in at less than \( D \). It can be seen that the parameters \( D \) and \( N \), and each \( d_i \), can be user-specified, or automatically determined.

Similarly, the most significant video skim of a recorded meeting can be based on the audio significance scores. The video skim can comprise the video clip produced based on the time index corresponding to the highest valued audio significance score in the manner discussed above. Where two or more scores have the same highest value, a resolution can be achieved in a manner as discussed above in connection with visual significance scores.

Alternatively, the video skim can comprise a plurality of video clips generated based on the audio scores. Thus, the first M highest audio scores can be selected, from which M time indices can be obtained. The video skim can comprise the plurality of video clips produced based on the time indices in the manner discussed above.

In still another variation, the most significant video skim of a recorded meeting can be composed based on visual and audio significance scores. In a simple composition, the video skim comprises a first video clip based on the highest value visual activity score and a second video clip based on the highest value audio score. Alternatively, the video skim can comprise a first plurality of video clips based on the N highest visual activity scores and second plurality of video clips based on the M highest audio scores.

It can be appreciated that video skims can be generated in various ways using the visual and audio scores. It can be further appreciated that the next most significant video skim
could comprise video clips based on the next highest visual activity score, or on the next N highest visual activity scores, or on the next highest audio score, or on the next M highest audio scores, or on both the visual and audio scores.

In accordance with another aspect of the present invention, a single metric can be computed by combining the video, audio, and textual scores to produce a composite score. For example, an arithmetic combination can be performed such as multiplication of the scores, addition of the scores, or some combination of multiplication and addition. Generally, it can be appreciated that some analytical treatment can be performed to produce the composite score.

If the composite score is computed for a meeting moment at time $t$, a meeting segment can be defined as the meeting recording between $[t-A_t, t+A_t]$. Segments from the recording corresponding to N such scores can then be selected and combined into a video skim of the meeting recording.

The video skims can be enhanced with additional video clips. The following discussion applies to each of the above-described processes for generating video skims. For example, consider when the most significant event is to be accessed from the recorded meeting. A video skim representing this can be produced in accordance with any one of a number of various embodiments of the invention. As another variation, the video skim can further comprise a video clip extracted from the recorded meeting based on the keyword audio segment metric (DO$_k$) having the highest value. The keyword audio segment and its corresponding video track can be included along with the other video clips which comprise the video skim.

From the foregoing, it can be appreciated that additional modes of information can be captured and incorporated into the process for identifying significant events. These additional modes of information can be captured by appropriate devices collectively represented in FIG. 1 as other capture devices 104$d$. For example, slide presentations can be captured and used to identify significant meeting events. Slide projection systems can be equipped with digitizing devices to capture the image including text. Personal computers can include slide production software which can store the slide contents for subsequent processing. If necessary, optical character recognition (OCR) techniques can be applied to the captured slide images to extract textual content from the slides. TF-IDF analysis can be performed on the text and associated with “events” from the slide presentation. Thus, for example, an event can be the time when the title slide is presented. Other such events can be the time when an outline of the presentation is presented, or when the conclusions are presented.

Events may include capture of audio information from speakers during the slide presentation. For example, when the presenter(s) introduce themselves this could be deemed a significant event. When a discussion ensues among the participants, this might signify a relevant event. Another event might be the slide that includes a discussion of the main topic. A significant event is likely to be the slide that the speaker spent the most time discussing. Likewise, a significant event might be the slide that resulted in the most discussion as determined based on the audio activity metric. In addition, these discussion events might involve speech capture of the discussion participants and subsequent transcription of the captured speech. The above-described analytical technique can be applied to the transcribed speech text to produce document occurrence scores DO$_k$, where each keyword $k$ is associated with the audio clip (keyword audio segment) having the highest DO$_k$ as described above.

The slide(s) presented during the time period spanned by that audio clip can be associated with that keyword.

Another mode of information can be information captured from a whiteboard. Again the text can be captured and transcribed by various whiteboard capture devices which are known and commercially available. Discussion-related events can be detected based on the audio activity score. Keywords from the discussions can be processed by capturing the discussion and converting the captured speech to text. In addition to its associated audio clip (keyword audio segment), each keyword $k$ can be further associated with the captured whiteboard information corresponding to the time period spanned by that audio clip. Other events might include noting when a person steps up to the whiteboard to use it. Such an event can be detected based on a video activity metric generated from the captured video recording.

More often than not, the mainstay of a meeting is paper. The agenda of a meeting is likely to be on paper. The topics of discussion are typically disseminated on paper, a copy provided for each participant. Notes jotted down during the meeting are recorded on paper. This represents a large source of useful text that can be subject to analysis, particularly notes produced during the meeting. The text contained in these papers, can be converted to text via OCR processing.

Events can be determined based on the note taking activity. For example, the time-spans can be recorded of when meeting participants take notes. An importance score can be assigned to a time-span based on the number of people taking notes, length of the notes, and so on.

Identification of meeting participants can be used as a basis for enhancing detected events. For example, the video of an event determined to be significant by any of the foregoing techniques can be further analyzed to identify the participants in the video. An importance score can be assigned to the current speaker based on the speaker’s rank in the company. Those meeting segments containing higher ranked employees may potentially be more important than other segments. The importance score can be based on speaker dominance. If one or several speakers dominate the meeting, the meeting segments containing those speakers are potentially more important than other segments. The importance score can also be based on the number of active speakers. The first or the last time a speaker speaks can be potentially more important than the rest of the meeting, and so the importance score can include that consideration.

Thus, in accordance with another implementation of the invention, the various modalities (video, audio, various forms of text) can be scored and used to identify significant meeting segments. For example, an importance score can be computed for each modality associated with each second of a meeting. The importance scores can then be combined to produce a significance score, for example by multiplying together all of the scores, or by summing all the scores, or by some other combination. The significance scores can then be ranked.

A video skim might comprise a number of 10-second segments of the meeting recording. For example, the $n^{th}$ most significant video skim of a meeting can be defined as a group of M segments, where the segments are based on the $n^{th}$ group of M consecutive significance scores taken from the ranked list. Thus, the meeting segments comprising the most significant video skim can be selected by taking 10-second segments of the meeting recording at those time indices having the M highest corresponding significance scores.

It was noted earlier that an embodiment of a meeting recorder in accordance with the present invention can be
portable device. A portable meeting recorder offers certain advantages, but presents its own unique challenges. Unlike a meeting recorder which might be based on instrumentation of a conference room where most meetings are carried out in one place, meetings recorded with a portable meeting recorder can take place in different locations. Identifying the meeting location can provide a very useful retrieval cue to facilitate identifying significant segments of a recording of a meeting that comprises many meetings taking place in various locations. One possible solution is to incorporate a GPS device into the portable meeting recorder. However, the accuracy of current GPS technology may not be sufficient to accurately identify the meeting locations, especially considering that they take place indoors.

Referring to FIG. 8A, our solution is based on recognizing a meeting room (or more generally a meeting location) from visual clues contained in the video data components of a meeting recording. We first perform background/foreground extraction as the recorder is manually operated and therefore it is unreasonable to assume that a clean shot of the background can be obtained with no person in the room. We use adaptive background modeling to extract the background. Our algorithm is based on an extension of a method discussed by Stauffer, C. and Grimson, W. E. L., Adaptive Background Mixture Models for Real-Time Tracking, Proceedings of Computer Vision and Pattern Recognition, pp. 246-252, 1999. A Gaussian mixture approximates the distribution of values at every pixel over time. For each Gaussian constituent, its likelihood of being background is estimated based on its variance, frequency of occurrence, color and neighborhood constraints. From this, an image of the background can be constructed based on the most likely background Gaussian at every pixel. Since this background estimate changes over time, for example due to the movement of objects in the room, we extract a new image every time a significant change in the background model is detected. These images are warped into a panoramic cylindrical projection as shown in FIG. 8A. The first image in the figure shows the meeting room with some background and foreground objects. The second image shows the foreground. The third image shows the background separated by our algorithm.

To identify the location, the background images can be matched against room templates in the database. Since the number of placements for the recorder in a particular room is usually limited, the templates can be categorized organized and stored as separate templates. In our case, one template is obtained from each end of a table in a conference room. We match the templates with the backgrounds of the meeting recordings by comparing their color histograms. The histograms are formed in the HSV color space because distance values in this space approximate human perception. The color space represented with 256 bins, where Hue is quantized into 16 bins, and Saturation and Value are quantized into 4 bins each.

Several background images are extracted for each meeting and an intersection histogram is computed using the histograms of these images. The intersection histogram is compared using Euclidean distance with each template in the database to find the closest matching meeting room. Employing an intersection histogram allows us to further eliminate the non-stationary objects in the meeting room and smooth out any background extraction errors. The use of multiple templates for each room provides a robust method for location identification. In our experiments, we successfully identified the 5 meeting rooms that we have in our research facility. Improvements to the algorithm might include using the size and the layout of the meeting room to address the issue of distinguishing rooms with similar colors.

Recorded meeting segments can be enhanced by further identifying and locating the meeting participants in the recorded segment. Locating meeting participants is a non-trivial problem, especially considering that a clean shot of the background is typically not available and participants are likely to have minimal motion. We address this problem by using sound localization to find the approximate location of each meeting participant. Then the precise location of each face is found by identifying the skin regions in this approximate location.

Skin pixels are detected in the normalized RG-space as discussed, for example, by Waibel, A., Bett, M., Metze, F., Ries, K., Schaal, T., Schultz, T., Soltau, H., Yu, H., and Zechnar, K., Advances in Automatic Meeting Record Creation and Access, Proceedings of the International Conference on Acoustics, Speech, and Signal Processing, 597-600, 2001. Small holes in skin-colored regions can be removed by a morphological closing and then connected component analysis can be used to identify face region candidates. In environments with complex backgrounds, many objects, such as wood, clothes, and walls, may have colors similar to skin. Therefore, further analysis of the skin-colored regions, using techniques such as orientation and geometric feature analysis can be applied to further eliminate non-face regions. Some example face localization results are shown in FIG. 8A.

One of our goals is to find representative shots of the meeting attendees that can be included in the meeting description document. It is possible to extract many shots of a participant from the video sequence. However, generally not all of these shots are presentable. It is desirable to obtain frames where the individual is not occluded and facing the camera.

An example of obtaining a “best shot” from a video sequence is shown in FIG. 9. First, the source video (in this case, the doughnut shaped video 902 from an omni-directional camera) is captured. Several still shots of the speaker are then extracted. In one implementation, sound localization information 912 is used to identify candidate still shots; e.g. a still from when she/he first starts speaking 904a, a still from when she/he finishes speaking (for the first time) 904b, and one between these two times 904c.

These shots are then evaluated to pick the best shot of a meeting participant. For example, the best shot 910 can be selected by evaluating the size of the face region relative to the size of the body region and/or the whole frame, computing the ratio of the face width and height, and evaluating the ratio of number of skin pixels detected in the best-fitted ellipse (to the face region) to the area of the best-fitted ellipse. FIG. 9 shows various regions 906 that can be considered. The larger faces with more skin pixels are selected as better shots.

In a specific implementation, high resolution video capture devices can be used to produce high resolution images of the participants. This provides sufficient resolution in the captured video for the application of computer image analytical techniques to identify certain facial features, such as eye and mouth regions. The selection of the best attendee shots can be based on the detail of the facial features, such as the mouth and eyes, the distances between these facial features and their relative locations, the size and geometry of the face region, and so on. When the image of the face of a person looking straight at the camera is taken, these parameters tend to fall within certain ranges. These ranges can be
the basis for computing templates to facilitate making a
determination whether a person is looking straight at the
camera.

FIG. 9A shows a schematic representation of a template,
highlighting some possible facial metrics that can be used
to create a template. For example, the template 902 comprises
a pair of eye regions, a nose region, and a mouth region.
Some metric exemplars of the template might include an eye
separation $D_n$ measured between the center of each eye. A
nose angle $\theta_n$ can be measured as the angle between two
lines passing through an interior of the nose region to the
center of each eye. A mouth separation $D_m$ can be the
distance from the eye to the mouth. Similarly, a nose
separation $D_n$ can be the distance from the eye to the nose.

In a face image of a meeting participant who is facing
away from the camera, the various facial features metrics
will deviate from the template metrics. For example, in a
face image of a person whose face is turned to one side, the
eye separation will be less than if the person had faced the
camera squarely due the effect of the image of the eyes being
projected onto a two dimensional surface. Thus, a best shot
determination from among many “shots” (images) of a
meeting participant’s face can be made, first by identifying
the facial features of interest for a given image. Next,
metrics among the facial features are determined, such as
shown in FIG. 9A. These determined metrics are then
compared against their corresponding template metrics. An
error computation can be performed to produce an error
quantity for that image. For example, each template metric
can have an associated tolerance to establish a tolerance
range. Each determined metric can be compared against the

corresponding template metric to determine it falls within or
outside of the tolerance range. If the determined metric falls
outside of the range, the amount of deviation can be
recorded. A composite error quantity can be computed based
on the number of determined metrics that fall within the
tolerance range and the amount of deviation of the deter-

mined metrics that fall outside of the tolerance range. The
process is repeated, and the image having the smallest error
quantity can be deemed to be the best shot of that person.

The locations of the eye pupils can be detected using an
IR emitter attached to the meeting recorder as an enhance-
ment to the camera. (For example, J. Davis and S. Vaks in
A Perceptual User Interface for Recognizing Head Gesture
Acknowledgements, ACM Workshop on Perceptual User
Interfaces, Orlando, Fla., Nov. 15-16, 2001, describe such a
system.) Sound localization can then be combined with the
IR-based eye pupil detection to eliminate false positives in
pupil detection. Recall that sound localization data includes
at each time index information indicating the direction of
the sound source. The images each can be associated with a
time index. Each image that is analyzed for pupil identi-
fication can be cross-referenced with the sound localization
data based on the time index associated with the image in
order to provide an additional decision point for pupil
identification.

For example, suppose an image is being analyzed (using
known computer image analysis algorithms) to identify
pupils. Suppose further that a positive determination has
been made, indicating the presence of pupils. On the one
hand, one can simply assume the determination to be correct
and continue on. On the other hand, the sound localization
data can be used as an additional test to add some confidence
to a positive determination. The time index associated with
the image is used to access the sound localization data
corresponding to that time index. If the sound localization
data does not indicate that any sound originated from the
direction of the camera at that time index, then such an
indication can be interpreted to mean that the positive
determination was a false positive, and the quest for eyeballs
would continue. On the other hand, if the sound localization
data indicated a sound originated from the direction of the
camera at that time index, then the positive determination
can be deemed to have been confirmed. Generally, employ-
ing the additional information of the IR data and the sound
localization data can increase the reliability of identifying
facial features, in addition to identifying pupils. Further-
more, it can be appreciated that an analogous process can
also be used to reinforce the confidence of the sound
localization results based on the pupil location data.

FIG. 9B is a schematic representation of an image 922
taken by a camera, illustrating the metrics in an alternative
embodiment for determining best shot. Here, pupil location/
orientation and facial orientation can be used to make the
determination. The pupils can be identified in an image in
the various manners discussed above. In addition, a head
outline 934 is also identified from the image. After the pupils
are detected, their location/orientation can be determined. In
this embodiment, pupil orientation is made relative to a
frame of reference R; for example, the edge of the image.
Pupil orientation can be defined as an angle $\alpha$ of a pupil line
912 passing through the identified pupils 932 relative to an
edge (e.g., 922a) of the image, although any edge can be
used, or any other constant point of reference such an
alignment mark 924 that can be superimposed on every
image at a constant location. Pupil location D can be based
on the location of the pupil(s) 932 relative to the detected
outline of the head 934. The orientation of the face (head)
can be based on a line of symmetry 914 vertically bisecting
the detected outline of the face relative to a reference; e.g.
the edge of the image. Here, orientation is shown as an
angular measurement $\beta$. It can be appreciated that the
orientation can be defined by other similar metrics.

Pupil location metrics can be used to indicate the degree
to which a person is facing directly at the camera and thus
serve as a basis for determining the best shot. In an image
of a person squarely facing the camera, the pupils 932 will
be generally symmetric with respect to the face outline 934.
For example, metrics such as pupil distances from the left
and right side of the face outline typically will be symmetric.
Thus, for example, the $D_1$ and $D_3$ measurements should
be close, and the $D_4$ and $D_2$ measurements should be close. It
can be appreciated that other such symmetry measurements
can be used. When the head is turned, these measurements
would no longer be symmetric due to the effect of projecting
a three dimensional object onto a two dimensional surface.

Pupil and face (head) orientation can then be used in
conjunction with pupil location to further facilitate deter-

mining best shot selection in the following manner. These
metrics can indicate the degree of rotation of the head. For
example, a person who has fallen asleep is not likely to be
facing the camera directly, but rather will be facing the table.
Thus, for the metrics shown in FIG. 9B, the vertical bisector
914 will have a $\beta$ of about 90°, as will the pupil line 912.
Thus, each candidate image of a person being considered for
best shot selection will be analyzed to determine the value
of these metrics. The candidate image having the most
symmetry in terms of pupil location and having $\alpha$ and $\beta$
angles that show the least total deviation from 90° will be
selected as the best shot image for that person.

FIG. 10 shows an illustrative example of a user interface
in accordance with the invention. Meeting recording seg-
mants and computed scores can be presented to the user via
a suitable GUI as shown in the figure. In this particular
A SECTION tag typically marks a period of time (e.g., a one second period). For each SECTION we can have more than one SPEAKER. Each SPEAKER contains the values shown. All the data for a SECTION are shown at the specified TIME on the time line. The LOC tag determines the position of the speaker marking on the display; typically “1" is the leftmost location in the display bar. VALUE can be “0" or "1" indicating whether the person is speaking or not. Alternatively, VALUE can be a real number indicative of the strength of the measure of activity. The ID tag represents a particular speaker and can be used to indicate a color value associated with that speaker. However, we typically assign colors to speakers automatically since this gives the user some control over the visualization.

The Muvie client parses the XML format described above and creates a speaker data structure that can be used to create the display shown in FIG. 10. The speaker data structure contains a linked list of <SECTION> objects. Each <SECTION> object contains a TIME and a linked list of <SPEAKER> objects. Each <SPEAKER> object contains a LOC (location), a <VALUE> and an <ID>. As noted above, the <VALUE> can either be binary ("0" or "1") or a real number indicating “strength” approximation, if a more accurate representation is desired. In the XML fragment shown, values are in binary format. When the XML file is parsed, it is determined how many <SPEAKER> locations to allocate for this video.

FIG. 11 is an enlarged portion of the timeline 1004, showing a graphical representation of a speaker transition channel positioned next to a text thumbnail and keyframe thumbnail graphic. The speaker transition channel graphic is vertically arranged. The speaker transition channel shows 4 subchannels s1-s4, each representing one speaker location in the meeting. A fifth subchannel s5 is shown, but no activity is indicated, suggesting that the speaker has not spoken during the segment of the timeline shown in FIG. 11. Using the speaker data structure, the speaker transitions can be mapped into appropriate grid locations. For instance, it can be seen that the s1 speaker starts the meeting followed by the s3 speaker. To map the data on the user interface, we convert the time associated with the <SECTION> object into a pixel location using the following parameters:

\[
\text{time in seconds} = \frac{\text{image height} \times \text{pixel per second}}{\text{seconds per pixel}}
\]

For example, suppose we have:

- height=786 pixels,
- duration=1800 seconds (½ hour)
- time=0.4367 (pixel per second) and
time=2.29 (seconds per pixel)

Then, when we have a <SECTION> object with a <TIME> stamp of say 356 seconds, we can plot the associated <SPEAKER> structures at location \(<\text{TIME}\times t_{\text{pix}}\>) or 356*0.4367=155 pixels vertically relative to a reference location. FIG. 12 shows a result of iterating through all of the <SECTION> objects and plotting all of the <SPEAKER> data. In this case, each vertical subchannel in the speaker content channel can be assigned a unique color to distinguish it from the other subchannels.

A slider graphic 1016 (FIG. 10) can be manipulated along the timeline 1004 to specify a time index. The location of the slider graphic can be mapped to one or more time indices as shown above. Using the time index(ices), a corresponding
video skim can be produced in the manner discussed above and presented in a presentation window 1022.

A representation for the visual activity channel can be graphically represented as a histogram along the time line 1004. It displays the visual activity scores for each unit of time (e.g., seconds, minutes) of a video sequence. The score is computed per Eqn. 2 and can be normalized for the display area. A slider window 1020 can be used to navigate the video based on the visual activity scores. The video and the rest of the metadata are automatically synchronized during the playback. The output of visual activity analysis is given to the Muvie Client in the following XML format:

```
<MUovie-SECTIONS>
  <SECTION>
    <STIME>0</STIME>
    <ETTIME>4</ETTIME>
    <VALUE>20</VALUE>
  </SECTION>
  <SECTION>
    <STIME>1</STIME>
    <ETTIME>1</ETTIME>
    <VALUE>20</VALUE>
  </SECTION>
  <SECTION>
    <STIME>11</STIME>
    <ETTIME>11</ETTIME>
    <VALUE>20</VALUE>
  </SECTION>
  ...
</MUovie-SECTIONS>
```

The SECTION tag typically marks a one second period. However, it can be extended to longer periods of time. The STIME tag determines the start time and the ETIME tag determines the end time when a particular activity score is valid.

The Muvie client parses the XML format described above and creates a data structure that can be used to create a representation for visual activity analysis shown in FIG. 10. The XML file contains a collection of <SECTION> objects where each <SECTION> contains a <STIME> (start time), a <ETIME> (end time), and a <VALUE>. The <STIME> and <ETIME> values are represented in seconds but could also be represented in milliseconds. At present we use the same start and stop time, but one could easily modify this to show <SECTION> tags which depicted time ranges, e.g., 4 second time periods. The VALUE represents the visual activity taking place at a given time during a video.

To plot this information on a user interface, as we did with the speaker transition information, a video channel is included in the timeline 1004 where we plot the visual activity as shown in FIG. 13. The histogram shown in FIG. 13 is plotted with the X-axis oriented vertically and the Y-axis oriented horizontally such that histogram bars with higher Y values represent more activity than those with lower Y values. Each <SECTION> tag contains a start and stop time. As noted above, we are using the same start and stop time, but one could easily modify this to show <SECTION> tags which depicted time ranges, e.g., 4 second time periods.

To map the data onto the user interface, we first determine the highest value represented in the data collection. We call this value max. Next we determine the width of the user interface which we call graphWidth (see FIG. 13). This value is predetermined by a default setting or by user configuration. We use the combination of the max value and the graphWidth to plot the values on the Y-axis. We use the tpx_m to plot the values along the X-axis since these values are based on time.

For example, if we have a <SECTION> object with a <TIME> value of 491 and a <VALUE> of 56, where graphWidth=50 and max=78, then, we have a new multiplier graph_m=max/graphWidth or 1.56. We divide the <VALUE> by graph_m to produce the width of the histogram bar for this <SECTION> object. Therefore, in the above example, we would plot the <SECTION> with <TIME>=491 and <VALUE>=56 at the vertical location (time) 491*.4367=214 with a width of 56/1.56=35. This process is conducted on each <SECTION> object to produce a timeline channel similar to the one shown in FIG. 14.

A representation for the audio activity channel is shown in 1004 that displays the audio activity scores for each minute (can be in seconds) of a video sequence. This value is computed according to Eqn. 4 and scaled for the display area. It is displayed, navigated, and formatted in the same way as the visual activity data described above.

Completing the description of to FIG. 10, the presentation window 1022 can be utilized to display additional modalities such as captured slide images, captured whiteboard images, and captured meeting notes. As shown in the figure, the presentation window can display perspective and panoramic views of the meeting recording. The foregoing techniques for obtaining meeting location and meeting participants can be incorporated in a meeting summary window 1024. Best shot selection can be used to provide the best image available of each participant.

Using this interface, the user can browse a meeting by reading the description page, listening only to the speakers that he is interested in, looking at the high-motion parts, searching for keywords in the transcription, looking at the presentation slides and whiteboard images, and so on. In this way, hours of meetings can be browsed in much less time. The user interface can also support editing of the video, which enables the user to efficiently communicate meeting documents with others.

FIG. 15 shows an implementation of another aspect of the present invention in which summaries can be represented. A summary create check box 1502, or other such graphic, can be presented to a user. The summary create check box when activated (by “clicking” on it, for example) can determine summary segments from the meeting recording and provide a visual representation. In the example shown in FIG. 15, summary windows 1504 highlight those portions of the timeline which correspond to the identified summary segments.

The meeting summaries can be generated based on individual or combinations of visual scores, audio scores, text importance scores, and importance score computed from slide presentations, whiteboard capture data, and notes in the manner discussed above. The user can specify which modalities to use when generating the summary. Using the time indices corresponding to the scores of the selected modality(ies), summary windows 1504 can be generated and displayed.

For example, the user can specify a one-minute long meeting summary based on audio and visual activity importance. Thus, the N highest visual scores and the M highest audio scores can be deemed to constitute a summary of the meeting. As discussed above, video skims corresponding to the time indices of these scores can be generated. However, rather than displaying video skims, the range of time indices corresponding to the generated video skims can be used to generate the summary windows 1504 shown in FIG. 15. This allows the user to browse the recorded meeting with the summary windows acting as annotations signifying the significant portions of the event (e.g., a meeting) that was
20 recorded. The highlighted sections thus can serve as a guide for meeting reviews, navigation, editing, sharing, printing, and generally making life easier for the reviewer.

In another example, the user may request a five-minute meeting summary based on slide importance. The slides that the presenter had spent the most time on might be considered important. Thus, the meeting summary might comprise one-minute meeting summary segments corresponding to five such slides. Again, the range of time indices corresponding to the presentation of those slides would be used to generate summary windows 1504.

As an alternative to displaying summary window graphics 1504, portions of text in the transcription window 1514 corresponding to the time indices can be highlighted. Similarly, corresponding keyframes 1512 can be highlighted. It can be appreciated that other similar forms of indicating the recorded meeting segments comprising the meeting summary are possible.

Selection of a summary window 1504 such as by clicking can cause a corresponding video skim to be played. Similarly, clicking on a highlighted portion of text or a highlighted keyframe can cause the corresponding video skim to be played. Alternatively, all the video skims comprising the meeting summary can be played sequentially by clicking on a “playback” button 1522. Navigation control through a video skim can be provided by “reverse” and “fast forward” buttons 1524, 1526 provided via the interface, allowing the user to watch only the highlighted parts, or skip to other sections of the meeting.

FIG. 16 illustrates another GUT exemplar. Here, a markup language known as SMIL (Synchronized Multimedia Integration Language) is a public domain standard that is well-suited for displaying the variety of media that can be present in a meeting recording. A SMIL-based interface can be driven by the Real player display engine from RealNetworks, Inc. The omni-directional image of a meeting recording can be dewarped as described above and presented to the Real player via a SMIL script and presented in a window 1632 in the Real player. Keyframes 1634 extracted from the video can be shown with their corresponding time indices (time stamps) by the Real player. Clicking on a keyframe can result in the video being rewound to the corresponding time index.

The interface shown in FIG. 16 can provide a meeting summary function as described in FIG. 15. A summary create graphic 1602 can serve to initiate computation of a meeting summary. Keyframes 1604 which correspond to the meeting segments deemed to constitute the meeting summary (as determined based on one or more user specified modalities) can be highlighted to indicate the location of the meeting summary segments. Clicking on a highlighted keyframe can reposition the video in the playback window 1632 to the time index corresponding to the “clicked” keyframe. Navigation buttons, such as playback 1622, rewind 1624, and fast-forward 1626, can be provided, in addition to other conventional navigation controls, to facilitate navigating the summary segments.

What is claimed is:
1. A method for retrieving content comprising multimedia information, the method comprising:
   receiving a request to retrieve a portion of said multimedia information;
   in response to said request, producing a first portion of said multimedia information and presenting said first portion of said multimedia information, producing a term-frequency inverse document frequency (TF-IDF) score for each word in a plurality words contained in a transcript of said multimedia information;
   producing a plurality of keywords by taking words from said plurality of words based on said TF-IDF scores; and
   for each keyword, producing a document occurrence score for a plurality of segments of said transcript and associating a segment having a highest document occurrence score with said each keyword, said segments corresponding to portions of said multimedia information,
   wherein producing said first portion includes incorporating portions of said multimedia information corresponding to one or more of said segments of said transcript,
   said producing a first portion of said multimedia information comprising:
   selecting a frame of video from a plurality of video frames, said selecting based on a visual significance score, said visual significance score indicating a responsiveness of said selected frame to said request;
   selecting a segment of audio of said multimedia information, said selecting based on an audio significance score, said audio significance score indicating a responsiveness of said selected segment of audio to said request, said segment of audio having a corresponding video clip;
   generating a video skim from said selected frame of video and said video clip; and
displaying said generated video skim.
2. The method of claim 1 wherein said transcript further comprises textual material from one or more additional sources comprising whiteboard input, slide presentation material, and transcribed notes.
3. The method of claim 1 wherein said video information is MPEG encoded video data, and said video frames comprises I-frames.
4. The method of claim 1 wherein said audio significance score is based on a measure of speaking activity.
5. The method of claim 1 further comprising:
determining a best shot image from among candidate images of a person’s face comprising, for each candidate image, steps of:
   identifying a plurality of facial features;
determining a plurality of measurements among said facial features; and
   comparing said measurements against template measurements to produce an error quantity, said template measurements representing a plurality of average measurements of facial features, wherein a candidate image having a minimal error quantity is selected as a best shot image, wherein one or more such best shot images of people can be displayed along with said video skim.
6. The method of claim 5 wherein said identifying a plurality of facial features includes:
determining a time index associated with said candidate image;
determining sound directions occurring during a period of time corresponding to said associated time index;
identifying pupils from said candidate image; and
if a positive determination of pupil identification occurs, confirming said positive determination based on said sound directions.
7. The method of claim 1 further comprising:
determining a best shot image from among candidate
images of a person's face comprising, for each candi­
date image, steps of:
identifying pupils;
identifying a head outline;
determining location metrics of the pupils relative to
said head outline;
determining pupil orientation relative to a reference;
and
determining head orientation relative to a reference,
wherein a best shot is selected from among said candidate
image based on said location metrics, said pupil orienta­
tion, and said head orientation,
wherein one or more of such best shot can be displayed
with said video skin.
8. The method of claim 7 wherein said identifying pupils
includes:
determining a time index associated with said candidate
image;
determining sound directions occurring during a period of
time corresponding to said associated time index;
identifying pupils from said candidate image; and
if a positive determination of pupil identification occurs,
confirming said positive determination based on said
sound directions.
9. A system for accessing content from multimedia data,
the system comprising:
a data component configured to provide said multimedia
data;
a scoring component operative to produce scores corre­
sponding to segments of said multimedia data, wherein
said scores include visual significance scores and audio
significance scores; and
a selection component operative to select segments of said
multimedia data based on their corresponding scores,
wherein said selection is performed in response to a
request for a portion of said multimedia data,
wherein each of said scores indicates a responsiveness of
the corresponding segment to the request,
wherein said scoring component is further operative to
produce a video skim incorporating portions of said
multimedia data corresponding to one or more of said
selected segments,
wherein said scoring component comprises a computer
program configured to perform operations comprising:
producing a term-frequency inverse document frequency
(TF-IDF) score for each word in a plurality words
contained in a transcript of said multimedia data;
producing a plurality of keywords by taking words from
said plurality of words based on said TF-IDF scores; and
for each keyword, producing a document occurrence
score for a plurality of segments of said transcript and
associating a segment having a highest document
occurrence score with said each keyword, said seg­
ments corresponding to portions of said multimedia
data.
10. The system of claim 9 wherein said scoring compo­
nent comprises a computer program configured to perform
operations comprising:
obtaining a first video frame from video data contained in
said multimedia data, said first video frame comprising a
plurality of first subframes;
obtaining a second video frame from said video data, said
second video frame comprising a plurality of second
subframes, each of said first subframes associated with
one of said second subframes;
for each of said first subframes computing a local lumi­
nance metric as a function of said first subframe and its
associated second subframe, thereby generating a plu­
arity of local luminance metrics associated with said
first video frame; and
associating a visual score with said first video frame, said
visual score based on one or more of said local lumin­
nance metrics,
wherein said foregoing steps are repeated for a plurality of
video frames in said video data to produce a plurality
of visual scores, each of said visual scores associated with
each of said video frames.
11. The system of claim 10 wherein said local luminance
metric is determined for successive I-frames.
12. The system of claim 9 wherein said multimedia data
comprises MPEG (motion picture experts group) video.
13. The system of claim 12 wherein the video is omni­
directional video.
14. The system of claim 9 wherein said multimedia data
comprises video compressed using a block-based compres­
sion format including MPEG-1, MPEG-2, MPEG-4, H.263,
H.263L, and JVT/H.26L.
15. The apparatus of claim 9 further including a display
component operable to display said segments of said mul­
timedia data, said display component including a computer
program for determining a best shot image from among a
plurality of images of a person's face, said computer pro­
gram configured to perform operations comprising, for each
image:
identifying a plurality of facial features;
determining a plurality of measurements among said
facial features; and
comparing said measurements against template measure­
ments to produce an error quantity, said template mea­
surements representing a plurality of average measure­
ments of facial features,
wherein an image having a minimal error quantity is
selected as a best shot image.
16. The apparatus of claim 15 wherein said identifying a
plurality of facial features includes:
determining a time index associated with said candidate
image;
determining sound directions occurring during a period of
time corresponding to said associated time index;
identifying pupils from said candidate image; and
if a positive determination of pupil identification occurs,
confirming said positive determination based on said
sound directions.
17. The apparatus of claim 9 further including a display
component operable to display said segments of said mul­
timedia data, said display component including a computer
program for determining a best shot image from among a
plurality of images of a person's face, said computer pro­
gram configured to perform operations comprising, for each
image:
identifying pupils and a head outline;
determining location metrics of said pupils relative to said
head outline;
determining pupil orientation relative to a reference; and
determining head orientation relative to a reference,
wherein a best shot is selected from among said images
based on said location metrics, said pupil orientation,
and said head orientation.