Aproposal for the description of audio in the cont

extofMPEG-7

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Abstract

Sound content description is one of the aims of the retrieval of audio, there are other sound content-b arobust set of descriptors and structures for putt sound. Spectral Modeling techniques provide one usa descriptions. In this paper we will introduce one p sound descriptors that can be derived from them in features of a structure for organizing the informat Scheme"). All of our current descriptors can be con description of music (musical forms and styles, rol indeed. The descriptors proposed are the result of of them we have devised automatic extraction proced generic one that, based on a hierarchical (and recu levels of detail, addressing both syntactic (struct ural)

MPEG-7 initiative. Although MPEG-7 focuses on inde xingand asedprocessingapplicationswaitingtobedevelope doncewehave ingthemintorelation, and for expressing semantic concernsabout ble framework for extracting and organizing sound c ontent articularapproachtospectralmodeling, then we wi llpresentsome order to develop sound descriptions, and we will di scuss the ion that can be derived from them (a so called "Des cription sidered low-ormid-level, thus we will not covert hehighlevel esofcharactersinamovie,etc.)whichisalsore levantinMPEG-7 asoundanalysisbasedonaspectralmodelingtechn ique,andforall ures. The Description Scheme we present is intended tobea

rsive in some places) structure, can describe sound at multiple ural)andsemantic(content)waysfordescribingso und.

1.Introduction

MPEG-7 is an standardization initiative of the Moti audio coding like MPEG-1, MPEG-2 and MPEG-4, is mea describe sound [1]. The main application for MPEG-7 of audio, or of any other media. Although contentb conventional tasks as audioed iting, music composin also open new possibilities for live music mixing a music commercial assessment and recommendation, age

Audio content extraction and managing can be achiev anyway non-exclusive means: there are traditional s ign scene analysis techniques, statistical techniques, labeling techniques and the sauri that can work syne rgis to describe and organize more thoroughly the conten MPEG-7 will not standardize the way to obtain these descriptions and the way for structuring them.

onPicturesExpertGroupthat, insteadoffocusing on is mea nt to be an standardization of the way to shouldbethecontent-basedindexing and retrieval asedaudiodescriptionscanprove extremely fruitfu g, sound effects selection, or video cueing, they c an nd DJing, sound signaturing for copyright protectio n, ge ntbasedTV scheduling, etc.

chiev ed by different, sometimes overlapping, and ignal processing techniques, computational auditory
etc. [2, 3, 4, 5, 6]. There are also manual keyword rgistically with automatic techniques in order to h elp
n t of audio material. The key issue here is that descriptions nor how to use them, but only the

Describing sound content involves using procedures, techniques, and data, that have been found and developed in different researchareas (i.e. signal processing, music cognition, artificial intelligenc e, etc.), in order to solve problems as *Sound seg regation* (components of sound mixes need to be identified in order to describe them separately [7]), *Segmentation* (time-localized abrupt changes in significative parameters of sound have to be detected and classified as diagnostic cues for understanding changes in the content of the sonic flow), *Sound event and source characterization* (detecting pitch notes and

durations, chords, expressive gestures as vibratoo speaker, are at the basis of describing the molecul Analysis and Music Analysis (interconnecting the individualized sonic elements environmentsintoaglobalandabstractpicture,w

Our work on sound analysis inside the SMS (Spectral sinusoidalmodelsingeneral[9], and the tools we measures that can be considered as being useful des presented in the next sectional though, as we inten the extraction methods. The descriptors we use can specific needs. Thus, we can consider an instantane differences between pairs of contiguous or separate longer time scales, and, associated with that avera values at such a longer time scale. As it has been them can be used as a basis for elaborating higherreferringtosoundsinsearch, classification, and

Complementarily, audiocontent descriptions need a together with the additional information that could section the characteristics of such a describehighlevelfeatures(i.e.searchformale micro-structural (search for timbres with similar v example). As we will see, the description of an aud multiple levels: we can describe it frame-by-frame, butalsocanbedoneatahigherscale, taking into a spectral region along a segment. We believe that the efficient exp representation of sound can yield descriptions of (multimediaapplications.

rlegatoplaying, or identifying the gender and age ofa arelements of a stream of sound), or Auditory Scene of complex sonic hereroles, functions, and relationships are define d). Modeling Synthesis) framework [8] or havedevelopedsofar, makepossible tomanageseve ral criptors for the content of sound. They will be dthispapertobeofbroadinterest, we will not p resent be computed in different flavors, depending on ous value for them, a variation value for expressin g

d frames, an average value, for describing content at ge, a variance value, expressing the variability of the shown in relevant literature [2, 10] combinations o f f level descriptions that approximate ordinary ways o transformationtasks.

wayforstructuringandhandlingthosedescriptors be derived from them; we will discuss in another descriptor scheme, considering that sound descriptions not only voices, or forguitarsolos), but also macro-struct uraland ariations in the partial amplitudes along time, for iofilewiththehelpofspectralmodelscanbedon eat with so-called instantaneous or low level descript ors. considerationatemporal segmentofafileorstream, or loitation of such a kaleidoskopical mainly musical) sound content that will be usable i n

2.Spectralmodelsforsounddescription

There are several sound analysis techniques that ca LPC, cochleograms...) one of them being the spectral a decomposition, SMS[11]. In this type of analysis, the sound, from which we derive a time-continuous r tracks that follow the harmonic (or inharmonic) str subtracted from the original sound, obtaining ares approximation techniques. Contrasting with other an such kind of representation is very intuitive from databaseor audio processing software, and can be m or descriptors in a hierarchical way, in order to p Thus, we start with basic, and sometimes not quite sinusoidal track number three"), but we can end wit thewaysweusetotalkaboutsound(ascanbe"as

n be used to obtain content descriptions (wavelets, nalysis based on a sinusoidal plus residual we compute the short-time Fourier analysis (STFT) o epresentation of the sound in the form of sinusoida ucture of the sound. This sinusoidal component is idual signal that can be modeled with different spectral alysis techniques, as for example wavelets or LPC, the point of view of a final end user of an audio adequitemoreeffectivebyderivingotherattribut es reserve the information available at the lowest lev els. semantic descriptions (as can be the "amplitude of hmid-andhigh-level descriptors closely related to harpattackwithalongsustainedvibrato").

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Low-leveldescriptorsofsoundcontent

The descriptors that we use constitute a set for a microstructure of a sound. In this set there are ve But, starting from them, there are also other usefu abstraction of the sound characteristics, and that calculatedateachanalysisframefromtheoutputo thatotherrelevantdescriptorscouldbeused, as f attackharmoniccoherence[14], and some of themwi themomentthisisourlist:

simple parameterization that accounts for the ry basic parameters like instantaneous frequency, amplitude and phase of each partial and the instantaneous spectral characteristics of the resid ual signal. l instantaneous attributes that give a higher level will be listed below. These attributes are easily fthebasicSMS analysis. We should also acknowledg orexampleodd/evenpartialsratio[12],tristimulu s[13], llprobablybeincorporated sooninour system. For

- Amplitudeofsinusoidalcomponent: sumoftheamplitudesofallharmonicsexpressedi ndB.
- Amplitudeofresidualcomponent: energyoftheresidualcomponentexpressedindB.

- Spectralshapeofthesinusoidalcomponent: envelopedescribedbytheamplitudesandfrequenci es oftheharmonics, oritsapproximation.
- **Spectral shape of the residual component:** approximation of the magnitude spectrum of the residualsound.
- Harmonic distortion: measure of the degree of deviation from perfect ha rmonic partials.
- **Noisiness:** measure of the amount of nonsinusoidal informatio npresent in the frame. It is computed by taking the ratio of residual amplitude versus to talamplitude.
- **Spectral centroid:** the midpoint of the energy distribution of the mag frame. It could be considered as the "balance point" of the spectrum.
- **Spectraltilt:** theslopeofthelinearregressionofthedatapoi ntsusedtorepresentthespectralshape ofthesinusoidalpart.

Besides the instantaneous values, it is also useful to have parameters that describe the time evolutio n of an attribute. We describe it with the difference sbetween frames.

Another important step towards a musically useful p into fragments that are homogeneous in terms of cer extract segment attributes that will give a summary into semantic categories corresponding to sound eve generalsegmentationprocessdividesamelodyinto steady state and a release region. Global attribute average variation of each of the instantaneous attr variation, average amplitude variation, or spectral meaningful to extract the average and variance of e global attributes such as time-varying rate and dep expressive elements the interested reader can consu when we segment sound, we are using low-level conte (when appropriately combined and interpreted) meani instantaneous measures, the segment description use ordertogetthe"globalpicture"alongitsduratio n.

As a final issue, it should be noted that the descr allowabigdegreeofoverlappingregardingitsdes in existing systems for audiovisual content analysi descriptionasMPEG-7 should accommodate this kind providers' proprietary software the effective use off

arameterization is the segmentation of a sound tain sound attributes. Then we can identify and of its content, and may allow to classify the segm ent nts or sound objects. One of the most obvious and notesandsilencesandtheneachnoteintoanattac k,a s that can characterize attacks and releases refer tothe ibutes, such as average fundamental frequency centroidtrajectory[15].Inthesteadystateregi onsitis achoftheinstantaneousattributesandcalculateo ther thof vibrato and tremolo. For more details on these It another paper from our group [16]. In summary, nt descriptors that can also help to characterize ngful segments of sound. Contrasting with s statistical measures such as mean and variances in

iptors used in our spectral modeling environment criptionpower. Thiskindofoverlaphasalsobeen used s [2, 10]. We believe that a standard for content of redundancy, leaving to the front-endorcontent fthedescriptor'sset.

Mid-levelandhigh-leveldescriptors

Describing sound at mid-level means determining sou betweeneventsandobjectscanbealittlebitcont row source, and that any kind of behavior of that obje ctiss a duration property. Although in a Schaefferian sen they can be grasped through an operation of "reduce more convenient to separate sources from their beha

Moreover, describing sound at what we consider "hig into formal structures that convey musical meanings structures can be found in the implication-realizat ion Lehrdal & Jackendoff [19], but other non-musicologi c As it is a matter clearly outside the current scope cognitive/musicological issue than an engineering on e concentrate inmid-level descriptions, asitissti llcurren

Regarding the mid-level descriptions of sound, the alongside Rosch [20] the "basic level" of categoriz could be the description intowords). At alower le va attack/steady state/release, and at a higher one we identification and representation [21] is one of th multimedia databases and it has given birth to an i humming" [22, 23]. Recent incorporation of rhythmic kindof systems [24].

sou nd events and objects. Although the distinction roversial, we will consider that a sound object is a sound ctisan event. Events develop in time, so all even se [17] all events can be considered objects becaus e dhearing", from a functional point of view it seem s viors, a swedo.

hlevel", means incorporating events and objects and roles for them. Examples of such a kind of ion by Narmour [18], or in the generative theory by cal structures can be also considered as high-level of automatic procedures, and much a ne, we will keep it aside. On the other hand we wil 1 llcurrentlyanactiveareaofbasicandappliedre search. mostobviousdescriptionis-atwhatwecouldcall ation-the description into notes (in case of speec hit velweneedtodecomposenotesintoenvelopestages as needtogroupnotesintophrasesandmelodies.Me lody e most important problems faced by content-based nteresting new search modality called "query by constraints will improve the performance of that

Identifying sound sources is another "basic level" nature. Although we can use the same set of descrip sources is equivalent to describing a perceptual pr dimension (i.e. not as unidimensional as loudness o that determine one timbral sensation or another [25 problem by noticing that two sounds generated by th pitch do not share the same attributes that were sh processesforidentifyingsoundsourcesarenotqui results [26] [27]. For these reasons, similarity ba diverse and controversial. Anyway, systems as Sound IRCAM[29]showthateffectivefeaturesandprocedu

Onceasourcehasbeenidentifiedwecanaskforat wehaveidentified a sound a sone of a pianothede isaverticaloragrand,aSteinwayoraYamaha,e the instrumental family. As recent work reveals [30 identification instead of trying to directly identi related to the source is the acoustic environment w systematicway[31].

categorization task, but a very different one in tors than for identifying events, describing sound operty: timbre. As timbre is a complex perceptual r pitch), there are several variables and dimension s]. We can quickly grasp the complexity of this e same sound source, but separated two octaves in ared when notes were very close. Thus, automatic teeffectiveyet, although there are scattered inte resting sed searches and indexings of sounds are still quit e Fish [28] or the Studio On Line developed at the resdoexistforquitewellsolvingthosetasks. tributeslikematerialsormaker(forexample,once scriptionshouldbecompleted with attributes like ifit tc.).Wecanalsoaskformoregeneralattributess uchas], it could be easier to start with this level of fythe specific sound source. Another kind of descr iption here sound is produced, that can also described in а

but also others that might be proposed, we have

ion Scheme", that is, a structure that specifies th

and other Description Schemes. The scheme started

ngeFormat)[32][33], anongoing proposal devised

rvisualschemesinthecontextMPEG-7.

sound laboratories, that was intended for the

n of multichannel, multi-source sounds. It is

n a hierarchical (and recursive in some places)

tingdifferentkindsofdependenciesandrelationsh

ndabilityitcanbeusedtodescribesoundingener

ific schemes), and it can take into account differe

tionscanbechangedasneededanddependingonthe

escription sub-schemes that are not compulsory at a

lyonelevel, oratallavailablelevels of descrip

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tion.

3.ASoundDescriptionScheme

As a way of organizing not only our own descriptors devised what in MPEG-7 jargon is called a "Descript semantics and the relationships between Descriptors as an extension of SDIF (Sound Description Intercha through the collaboration of IRCAM, CNMAT and IUA storageofspectralanalysisdata.Sincethenitha sevolvedintoathoroughschemeinspirednotonly knowledgeaboutsoundbutalsoonproposalsmadefo

The scheme we present here addresses the descriptio intended to be a generic description scheme based o structure, that can describe sound at multiple leve lsofdetail(fromthelowlevelofanFFT analysis tothehighlevelofawholesoundfile),accommoda amongitscomponents.Givenitsmodularityandexpa (although speech description will require more spec kindsofdescriptorsofsound. The scope of descrip target application, and there are descriptors and d Thisway, the scheme can contain descriptions at on

The scheme has two main levels: syntactic and semantic. We need a syntactic level in order to describe the sound file or stream in a superficial, temporal-structural way. But we also need a semant ic levelinordertoassignsemanticlabelstotheele ments that deserve attention. The syntactic leveli sused to specify physical structures and the signal prope rties of the sound program. The elements of the syntacticpartoftheschemeare tracks, segments, specificinstances of segments like frames, and regions. On the other side, the semantic level is used to sp ecifysemanticfeaturesofasoundprograminterms of audioevents and audioobjects.

Otherelements at its topmost level are: a Model DS in a compact and abstract way, a Syntactic-Semantic elements of both levels, a Media DS for describing characteristics, MetaDS formanual descriptions, a ofrelevantinformation.

fordescribingtheanalysis and classification dat а Link DS for describing relationships between respectively storage format and other technical ndaSummaryDSforquickaudioandvisualbrowsing

DescriptionSchemesforthesyntacticlevel

TrackDescriptionScheme

AfirstsubstantialdifferencebetweenVideomedia several channels at the same time (stereo recording andSoundmediaisthatsoundcanbeexpressedalon g , Dolby Surround, multi-track recordings...). Those

channels may contain related information but also v to describe each track separately (while allowing)

ery distinctionesso it is clear that we need a sch eme inksbetween track descriptions).

A track has no more the length of the whole audio p not need to start at the same time, although it cou one track is needed. In case of being multichannel, situations where only one track could be used with mono channels are stored in one stere of ile, or whe practically identical...).

b rogram, although it could be shorter. Tracks do ldbe usual to do so. In case the program is monoo nly one track is provided for every audio channel. Cer tain multichannel audio could be envisioned (when double n content in two or more channels is supposed to be

AtrackDSisthendescribedby:

- a **TimeDS**, that describes the beginning and the end of the the rack (time relative to an absolute time of the whole file).
- a **Placement DS**, which describes the spatial placement of the trac k from the listener point of listening(i.e.left,right,60degreesleft,behin d,etc.)forthetrack.
- a **Segment DS**, which contains the different segments created by applying different segmentation criteriatothetrack, alongside additional segment ation information
- a **RegionDS**, that describes the microstructure of the track
- a **Track Linking DS**, that describes if the track is linked to another track (i.e. shares the same content descriptions) or any other kind of interact ion or relationship between tracks. When tracks are linked they share the same Segment DS (for exam ple stereo-recordings where both channels are related do not need separate Segment DSs. In th description of the left Track for the right Track)
- a **SummaryDS**, thatdescribesdifferentwaysforquicklistening or "auralization" of contents, and alsoforfastbrowsingofaudiocontents (asmusica lscore, midifile, spectrogram...)
- a **Media information DS**, which describes information specific to the stora ge media (i.e. samplingrate, resolution, format, compression form at, etc.)
- a Meta information DS, which contains information that usually cannot be extracted from the signalitself(title,author,technicalcrew,date...)

NotethatthelastthreeDS'sdonotappearinFigu re1atthetracklevelforkeepingthegraphmore clear.



Fig.1. OverviewofAudioDescriptionScheme

SegmentDescriptionScheme

An audio program can be segmented into one or more common property (for example, utterances of the mai vibrato note of a soprano...). Each one of such gro further segmented into another group of segments, y segmenttrees in order to accommodate different ove (i.e. by speaker gender and by background music/not

groups of contiguous samples that share some ncharacter of a video film, the chorus of a song, a ups is an audio segment. Audio segments can be ielding a tree segment. A track can contain differe nt rlapping segmentations according to different crite background music). General hierarchy of the Segment Tree goes from a root segment into subsegme nts, sub-subsegments, and finally into frames, which consist of samples.

Insummary, the Segment DS is described by:

- a TimeDS, that describes the beginning and the end of thes egment
- a MediainformationDS, a MetainformationDS, an AuralizationDS, and a VisualizationDS, • asdefinedinthetrackDS
- aSegmentationCriterion ,withthenameforthecriterionusedinthesegme ntation
- aSegment-RegionLinkDS, allowingtolinkthesegmentwithregiondescript ions(spectrallowlevelfeatures)
- several optional **Temporal Descriptors**, that describe temporal low-level features of the sound (i.e.autocorrelation,zero-crossingrate,etc.)

Lowestlevelsofasegmenttreecanbe:

the **Framelevel** : a frame is usually the shorter meaningful Segment DS. Itallows a deep description of the local sound characteristics through spectral an alysis (or other different technique). A frame has exactly the same properties than other Segment DSs. As it is the lowest level of the Segment DS the descriptors attached to this level representing tantaneousvalues.

n special cases like glitches, they do the Samplelevel : samples are like atoms of the sound, and except i not carry any meaning or content by themselves. Nei ther further decomposition of them is possible nor ngthislevelisthatsometimescanbeusefultoha linkingthemtoaRegionDS.Thereasonforincludi vea sample-by-samplesegmentdescription.

RegionDescriptionScheme

Itdescribesthemicrostructureofasegmentofsou spectrainside the effective bandwidth of a sound. long one or a frame-level one). That opens the poss microstructural level (when the Region DS is linked level (when the Region DS is linked to a higher-lev decompositions associated to the same segment, and sameregiondecomposition.

nd, in the form of a sound spectrum, or a series o fsub-Theregion can be linked with any kind of segment(а ibility for a description of sound ranging from the to a frame-level Segment) to the macrostructural el Segment). There can also be several region it is also possible that several segments share the

ARegionDSisdescribedby:

- a SpectrumDS , which describes the main spectral features of th esound
- anoptional Motion/DeformationDS, which describes the evolution of the Audio Spectr umalong ashortperiodoftime

SpectrumDescriptionScheme

ASpectrumDSconsistsof:

- a series of **Specific Global Descriptors** like *amplitude*, *fundamental frequency*, *spectral range*, noisiness, etc. The value for this descriptors is
 - \triangleright an instantaneous one when the Region DS to which it is connected is linked to the frame leveloftheSegmentTree
 - a **difference** one when the Region DS to which it is connected is \triangleright linked via a Region Motion/DeformationDStotheframeleveloftheSeg mentTree
 - a mean one when the Region DS to which it is connected is linked to a higher level of the SegmentTree
 - \geq a variance one when the Region DS to which it is connected is linked via a Region Motion/DeformationDStoahigherleveloftheSegm entTree
- a **SpectralShapeDS**, that describes the energy profile across the spec trumoftheframebothina global way with a Spectral Envelope ProfileDS (containing an envelope, LPC coefficients, Melcepstrumcoefficients, formants, etc) but also with optional descriptors used to describe specific features of the spectral shape (most of them were l isted in the low-level descriptors section, as for example: Spectral Centroid, Spectral Tilt, Noise Sh ape Harmonic Distortion, Odd/Even ratio, etc.).

DescriptionSchemesfortheSemanticlevel

In the semantic part we describe two broad categori temporal nature of audio events, they are mainly li Likewise audio objects are closely (but not exclusi be considered as temporal constrained behaviors of events.

es of sonic elements: events and objects. Given the nked with segments described in the syntactic part. vely)linkedtotheregionsdescribedthere. Events objects. Objects, then, are the generators of the

EventDescriptionScheme

An event is the temporal behavior of some audio obj ect along or around a certain segment of time. Typical audio events can be: a melody, a musical ph rase, a "solo" section, a musical note, different sectionsofanoteregardingitsamplitudeevolutio n(attack/steadystate/release),anaudiofade(in orout), a sentence uttered by somebody in a video, the word s in the previous sentence, the phonemes that the previous words are made of, non-linguistic utter and es(crying, shouting, sighing, etc.)...AnEventDS can contain an arbitrary number of Event DS. Therefore Event DS form a Tree (for example: a musical motive or melody is composed by different phrases, these phrases are composed by different notes, and the notes have different envelope sections). Of cou rse, there can be more than one tree. Therefore, in the EventTree, we won't talk about" violin" or " cello" (which are objects) but about their temporal beha vior suchas" violinphrases ",or" cellonotes ".

AnEventDSisdescribedby:

- an optional **Annotation DS**, which is a text descriptor for describing non-aut omaticallyextractablefeatures
- an **Event Division Criterion DS**, which describes the criterion used to divide the Event in Sub-Events
- an Event-SegmentLinkDS, whichlinksoneEventwithoneorseveralSegments
- asetof **TemporalDescriptors**

ObjectDescriptionScheme

The main function of an Object DS is describing sou nd sources. It is possible to distinguish different levels for describing audio objects. The most gener ic source objects can be: musical instrument, speec h voice and environmental sound/sound effects. For mu sical instruments more detailed levels can be: musical family of the instrument, specific instrume nt, excitation resonance characteristics (type and structure of excitation, resonance structure), mate rial that is made of, specific shape characteristic s. manufacturer and model, acoustic environment where they sound, etc. For speech voice: gender, age segment (child, young, mature, old), excitation-res onance characteristics, identity of speaker, acoust ic environment where it sounds, etc. Finally, for envi ronmental sounds: space/time trajectory of the sour ce, excitationresonancecharacteristics, specificsour ce.etc.

An Object DS contains an arbitrary number of Object DS and therefore form a tree (i.e. a musical ensemble can be an object made of groups of instrum these groups can be, in turn, made of small subgrou ps-strings: violins, cellos, violas; voices: sopra nos, tenors, etc.-, downtotheindividualinstrument, i fnecessary).

AnObjectDSisdescribedby:

- Anoptional AnnotationDS withtextdescriptionsextractedbyhand
- An **ObjectDivisionDS**, which describes the criteria used to construct the eobject tree
- Oneormore **ObjectTypeDescriptors**
- Oneormore **ObjectBehaviorDS** which describe the object behaviors (they link to events)
- **ObjectInteractionDS** which describes interactions between different obj ects (i.e. the bow and the body of a cello, a singer's voice and hermicro phone...)
- anoptional Object-RegionLinkDS ,forlinkingtheobjectwithitsrelevantspectral information
- anoptional Object-EventLinkDS .
- **ObjectDescriptors** which describe the sound sources at different level sas discussed above.

4.Conclusions

Spectralmodelsaresuitablefordescribingsounda tdi canbeusedasbuildingblocksformid-levelandhi gh derivethosedescriptions.Startingfromspectralm odd description scheme for audio in MPEG-7 that mainly descriptors, but can also accommodate high-level de scope of this discussion because they should be han view, MPEG-7 is a challenging initiative that shoul contents.Butitisalsochallengingfromanacadem ic problems that are still hot research topics for the ordertoprovideause ful and long-lifest and ardfo

tdifferentlevelsofabstraction.Lowleveldescri ptions gh-leveldescriptions, or formodels[34]thatallo wto odels representational framework we have developed a ainly encompasses low-level and mid-level audio e scriptors (although they have been kept outside the dled in a non automatic way). From a user point of d improve our efficiency for accessing multimedia ic and engineering point of view because it address es audio community, and therefore they must be solved in rmultimedia content description.

References

Seculmenting March (1000)
[2]Wold,E.,Blum,T.,Keslar,D.,andWheaton,J. :Content-basedclassification,search,andretriev alofaudio.IEEE Multimedia (1996)27-36
 [3] L. Wyse, Smoliar, S.W.: Toward content-based au technique. In: Rosenthal, D.F., Okuno, H.G. (eds.) (dio indexing and retrieval and a new speaker discrimination of the provide technique of techniq
IJCAI-95Workshop.Erlbaum(1998) [4] Martin, K. D.: Toward automatic sound source re cognition: identifying musical instruments. In: Pro c. NATC
ComputationalHearingAdvancedStudyInstitute,II Ciocco,Italy,July1-12(1998) [5]Foote,J.T.:Asimilaritymeasureforaudiocl assification.In:Proc.AAAI1997SpringSymposium onIntelligen
IntegrationandUseofText,Image,Video,andAudi oCorpora,Stanford,CA,March(1997) [6]Zhang T andKuo C L Content-basedClassifi cationandRetrievalofAudio In:SPIE's43rdAnnu alMeeting_
Conf.on Advanced Signal Processing Algorithms, Arc hitectures, and Implementations VIII, SPIE Vol. 3461 , Sar Diego, July (1998) 432-443,
[7] Scheirer, E.D.: Towards music understanding wit hout separation: segmenting music with correlogram
[8]Serra,X.,Bonada,J.:SoundTransformationson theSMSHighLevelAttributes.In: Proc.DAFX98:FirstDigital AudioEffectsWorkshop.Barcelona(1998)
[9]Peeters,G.,Rodet,X.:SignalCharacterization intermsofSinusoidalandNon-SinusoidalComponen ts.In: Proc DAFX98:FirstDigitalAudioEffectsWorkshop.Barce lona(1998)
[10] Pfeiffer, S., Fischer, S., Effelsberg, W.: Au tomatic audio content analysis. In: Proc. ACM Multi media 96 Boston, MA.November (1996)21-30
[11] Serra, X.: A system for sound analysis/transfor rmation/synthesis based on a deterministic plus sto chasti
[12]Kostek, B., Wieczorkowska, A.: Parametric repr esentation of musical sounds. Archives of Acoustic s. 22(1).3- 26 (1997)
[13]Pollard, H.F. Janson, E.V.: Atristimulusmeth odforthespecificationofmusical timbre. Acustic a, 51(1982)
[14] Grey, J. M.: Multidimensional perceptual scali ng of musical timbres. J. of the Acoust. Soc. of Am erica, 61 (1977)1270-1277
[15]Hadja,J.,Kendall,R.,Carterette,E.,Harshb erger,M:Methodologicalissuesintimbreresearc h.In:Deliège,I.,
Sloboda, J. (eds.) Perception and Cognition of Nusion C. EastEssex: Psychology Press (1997) [16] Herrera P and Bonada I. Vibrato extraction and parametrization in the Spectral Modeling Synth esi
framework.In:Proc.DAFX98:FirstDigitalAudioEf fectsWorkshop.Barcelona(1998)
 [17]Schaeffer,P.:Traitédesobjetsmusicaux.2èm [18]Narmour,E.:TheAnalysisandCognitionofMel UniversityofChicanoPress(1992) eédition.Paris.ÉditionsdeSeouil(1977) odicComplexity:TheImplication-Realizationmodel. Chicago
[19]Lehrdahl,F.,Jackendoff,R.:Agenerativethe oryoftonalmusic.MIT,Cambridge,MA(1983)
[20] Rosch, E.: Principles of Categorization. In: E . Rosch and B.B. Lloyd (Eds.), Cognition and Catego rization Hillsdale NI-Lawrence Erlbaum (1978)
[21]Lindsay,A.T.:Usingcontourasamid-levelre presentationofmelody.MScthesis.MIT,Cambridge, MA(1996)
[22]Ghias, A., Logan, J., Chamberlin, D., Smith, B. C.: Querybyhumming: musical information retrieva lina naudio database. In: ACMMultimedia, 95-Electronic proceedings. San Francisco, CA, Nove mber (1995)
 [23] McNab, R.J., Smith, L.A., Witten, I.H., Hender son, C.L., Cunningham, S.J.: Towardsthedigitalmu siclibrary tuneretrieval from acoustic input. In: Proc. ACMD igital Libraries 96 March (1996)
[24] Handel, S.: Timbre perception and auditory obj ectidentification. In: Moore, B.C.J. (ed.). Hearin g.San Diego, CA: Academic Press (1995)
[25] Shmulevich, I., Yli-Harja, O., Coyle, E., Pove l, D., Lemström, K.: Perceptual issues in music pat ter
recognition-Complexityofrhythmandkeyfinding.I n:Proc.ofSymposiumofArtificialIntelligencea ndMusica Creativity.Edinburgh(1999)
[26] Brown, J. C.: Computer identification of music al instruments using a pattern recognition with cep stra coefficientsasfeatures. Journalofthe Audio Engi neering Society, 105(3) (1999) 1933-194
[27]Martin,K.D.:Sound-sourcerecognition:atheo ryandcomputationalmodel.Ph.D.Thesis,MIT.(1 999)
[28] Blum, T., D. Keislar, J. Wheaton, and E. Wold .: Audio databases with content-based retrieval. In :Maybury M.T.(ed.)IntelligentMultimediaInformationRetri eval.MenloPark, CA:AAAI/MITPress(1997)
$[29] Ir cam-SOLS earch Engine page: http://www.ir cam .fr/equipes/instruments/sol_psy/Scripts_english/cuidad4.html .fr/equipes/instruments/sol_psy/Script$
[30]Martin,K.D.andKim,Y.E.:Musicalinstrument identification:Apattern-recognitionapproach.In :Proc.ofthe AcousticalSocietyofAmerica.October(1998)
[31]Jullien, J.P., Kahle, E., Winsberg, S., Warusf el, O.: Someresults on the objective and perceptual characterization
[32] Wright, M. A. Chaudary, Freed, A., Wessel, D., Rodet, X., Virolle, D., Woehrmann, R., Serra, X.: N ev
applicationsoftneSoundDescriptionInterchangeF ormat.In:Proc.ICMC-98,AnnArbor,MI(1998) [33]Ircam-SDIFWebpage: http://www.ircam.fr/produits/techno/multimedia/Cui/dad/SDIF-e.html
[34]Casey,M.A.:AuditoryGroupTheorywithApplic ationstoStatisticalBasisMethodsforStructured Audio.Ph.D Thesis.MITMediaLab(1998)