



Unleashing Video Search

**John R. Smith (with Apostol Natsev, Quoc-Bao Nguyen, Jelena Tesic,
Lexing Xie, Rong Yan, Joachim Siedl, Christian Penz, Florian Letz)
Senior Manager, Intelligent Information Management Dept.
IBM T. J. Watson Research Center
Contact: jrsmith@watson.ibm.com**

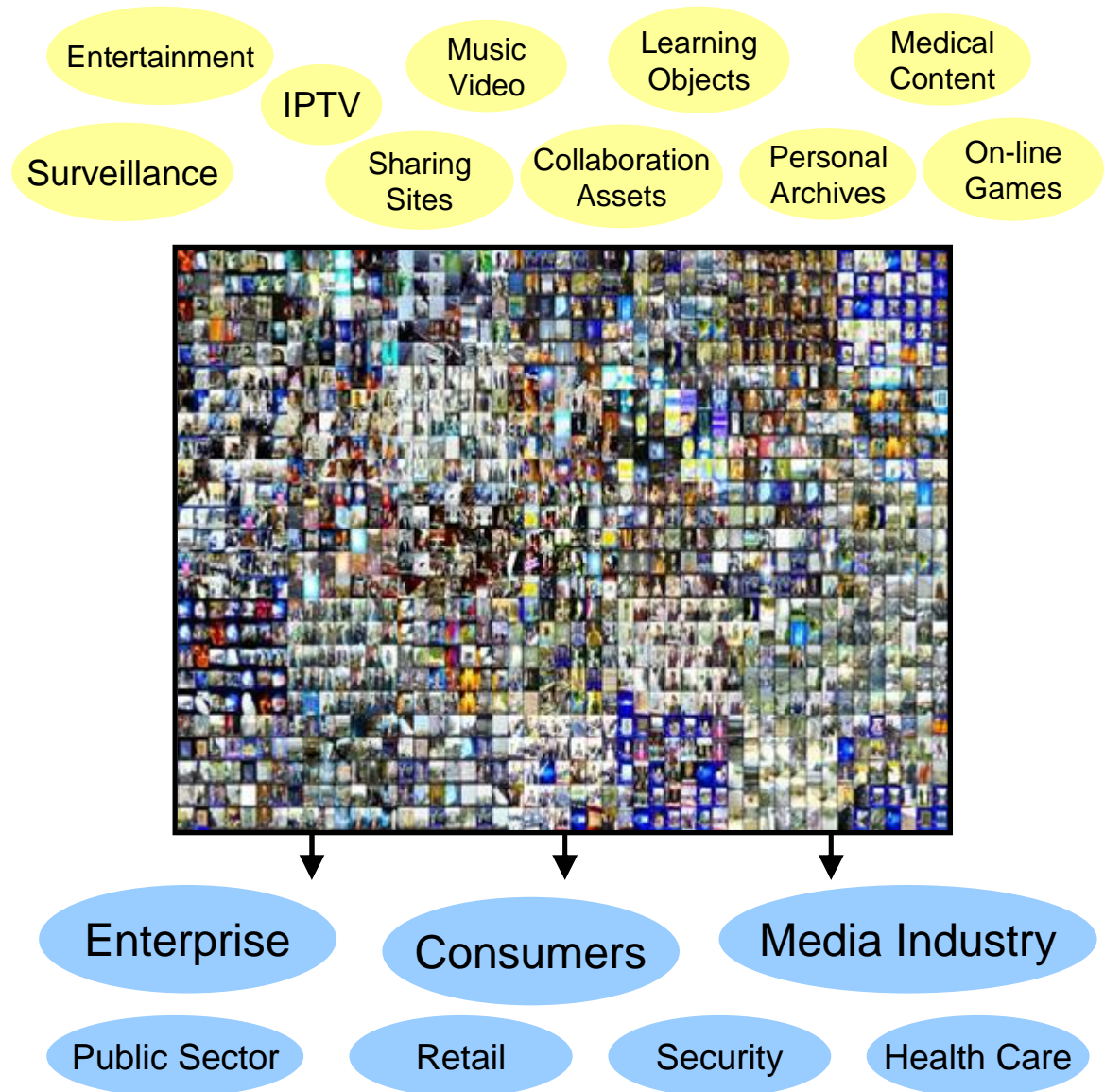
May 2008

Outline

- Challenges of video search
- Making video search better:
 - Visual recognition of content
 - Semantic labeling of visual clusters
 - Multi-modal video search
 - Concept-based video query expansion
- Video retrieval evaluations:
 - TRECVID
 - VideOlympics
- MPEG-7 multimedia content description standard

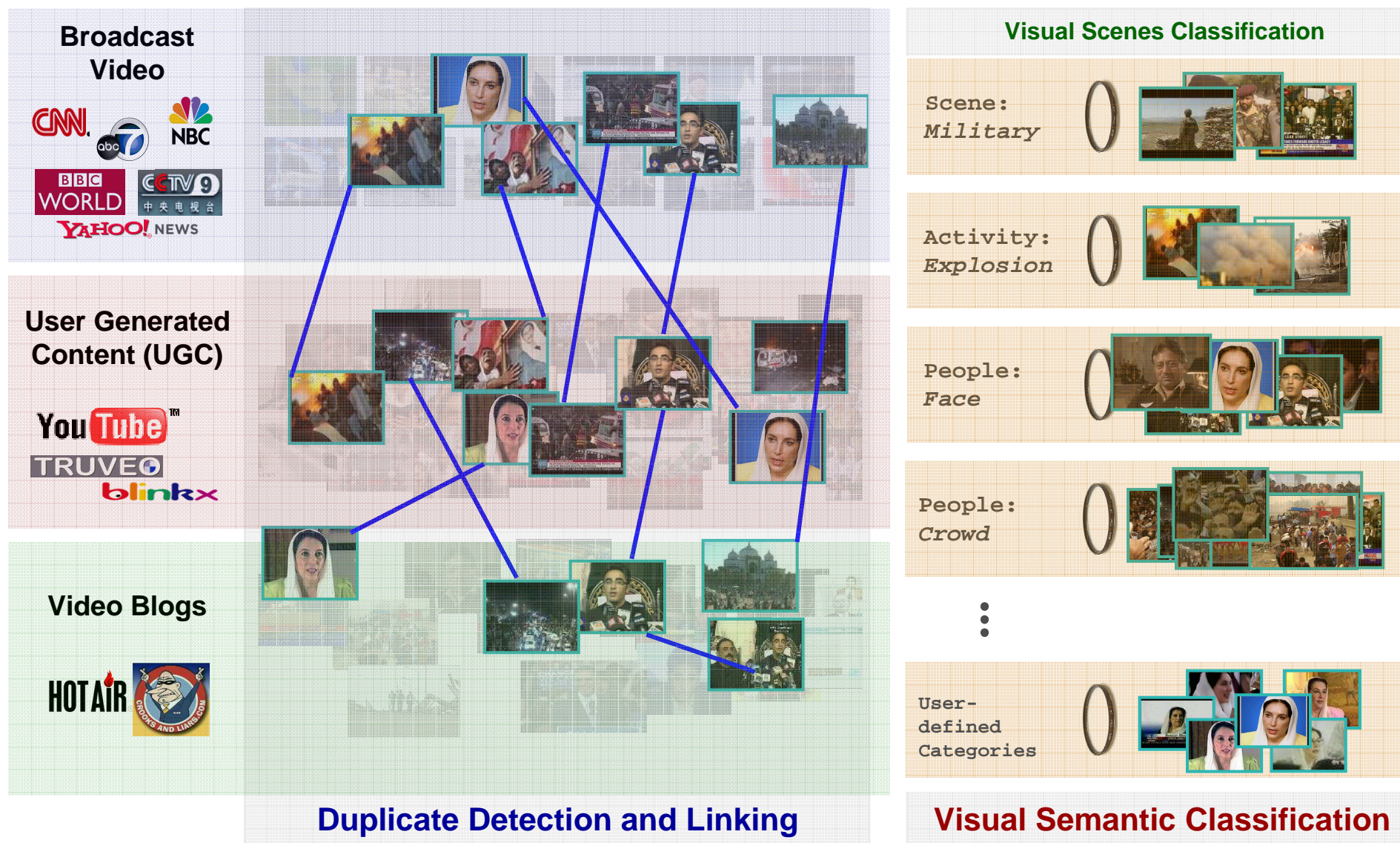
Video is rapidly becoming a regular part of our digital lives

- Growing deluge requires more effective solutions for organizing, managing & searching video content
- Manual indexing is costly, time-consuming and inadequate
- New technologies are needed to automate processing and unlock value of large repositories
- Metadata standards are needed to support interoperable search

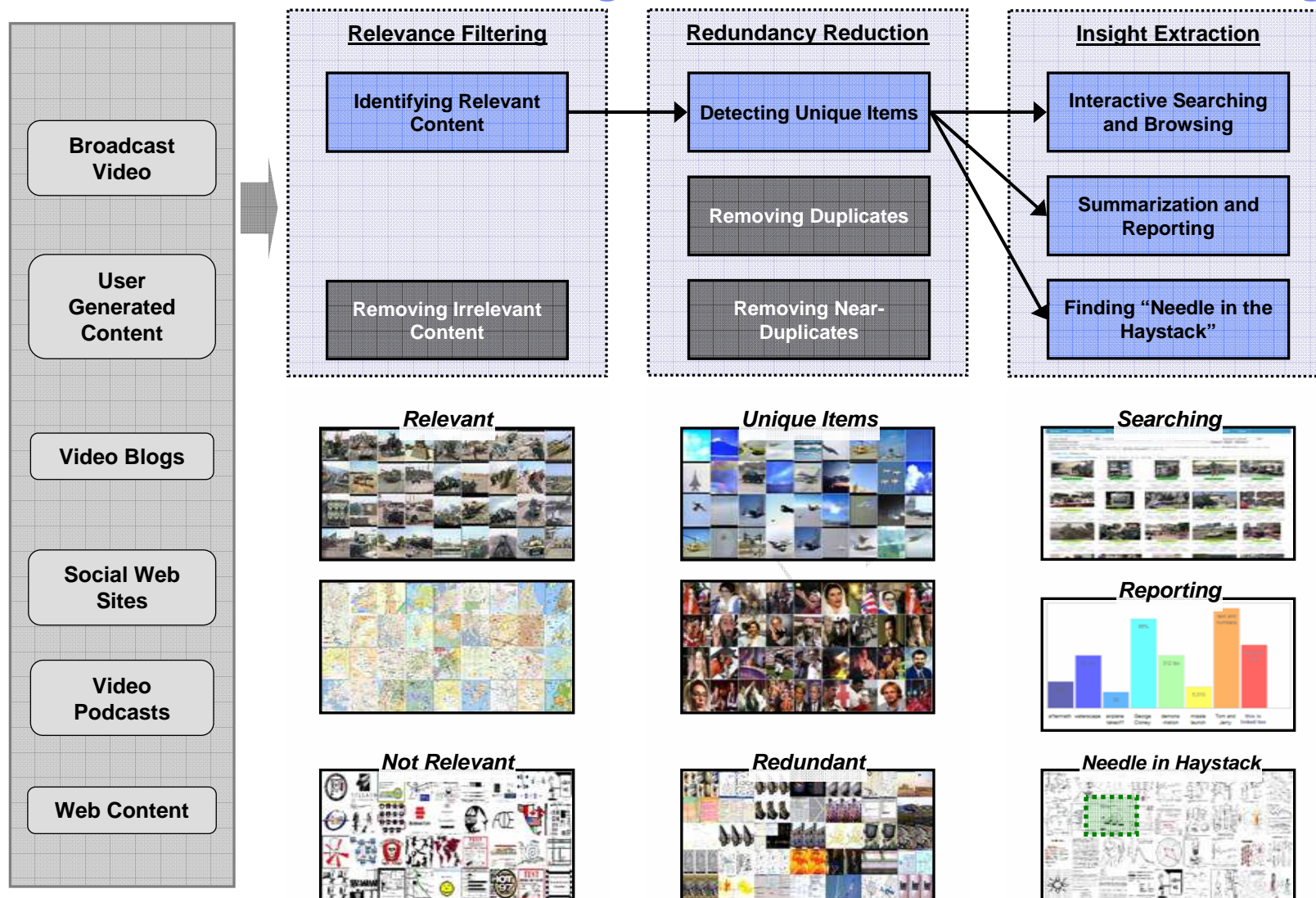


Tremendous growth of video is increasing expectation that it will be as easy to search as text

Insight Extraction Across Diverse Sources of Video and Image Content



Ability to process and recognize visual semantics in video & image data can turn massive amounts of digital content into actionable intelligence

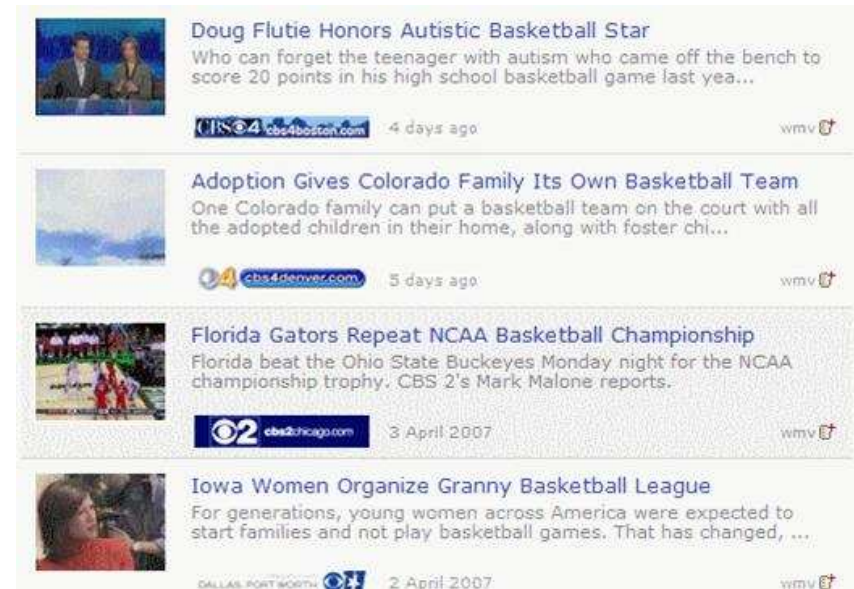


Unfortunately, it is still difficult to find relevant video content

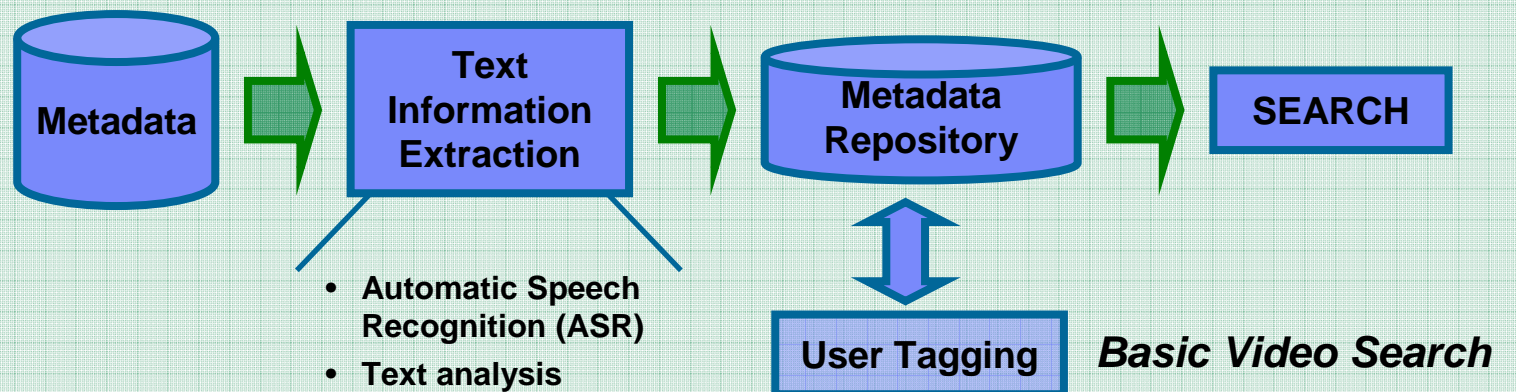
Today's Basic Video Search is not Satisfying for Users

- Frustrating: too many videos to wade through
- Chaotic: hard to find content of interest
- Funky: cannot separate professional from UGC
- Inconsistent: video quality mixed

*www.emarketer.com



- Program guides (EPG)
- Professional metadata
- Web text
- Audio transcripts



Today's Web-based video search is not adequate in either depth or breadth

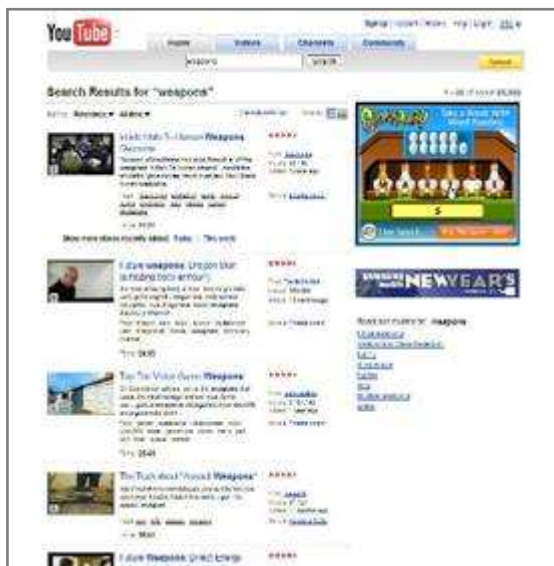
YouTube — “Weapons”

Depth:

- Cannot distinguish matches showing weapons scenes

Breadth:

- Does not broadly search the Web
- User-generated and user provided video



Blinkx — “Weapons”

Depth:

- 53,000 matches related to “weapons”
- No way to obtain clips showing weapons scenes

Breadth:

- Results limited to partner content



Clipblast — “Weapons”

Depth:

- Search relies on text descriptions

Breadth:

- Limited to partner content



TruVeo (AOL) — “Weapons”

Depth:

- No ability to refine search based on visual content
- Search relies on text scraping from Web

Breadth:

- Preference for AOL and partner content



All video search relies on metadata (e.g., manually authored, automatically extracted, scraped, etc.) – but, today's metadata is not good enough!!!

Issue	What's wrong
Too sparse	Few video objects have any metadata
Inadequate	Mainly tags or few keywords, program-guide info for broadcast video, speech available in few cases
Coarse-grain	At level of digital objects only
Not visual	Does not describe what is visually depicted
Ambiguous	Taxonomies not widely used; folksonomies creating new problems
Inconsistent	Vocabularies and taxonomies not standardized
Subjective	Limited verification across users
Not trustworthy	Professional metadata mixed-in with noise

Professional Cataloging and Social Tagging Approaches

IBM Research

Manual Cataloging – By Professionals

Pros	Cons
<ul style="list-style-type: none"> Controlled vocabularies & standard taxonomies Higher quality 	<ul style="list-style-type: none"> Costly Human resource intensive Cannot keep up
<ul style="list-style-type: none"> Example: Fox, CNN, BBC, Broadcast TV 	

Automated Tagging – By Machine

Pros	Cons
<ul style="list-style-type: none"> Lower human cost Domain & data driven approach to semantics 	<ul style="list-style-type: none"> Requires training of models Lower quality than manual tagging
<ul style="list-style-type: none"> Example: Marvel, Informedia, TRECVID concept detection 	



Social Tagging – By Users

Pros	Cons
<ul style="list-style-type: none"> User driven Emergent folksonomies Serendipitous browsing 	<ul style="list-style-type: none"> Ambiguity Uncontrolled vocabulary Synonyms
<ul style="list-style-type: none"> Examples: Delicio.us and Flickr 	

Popularity

High-value content, hit-TV shows, movies

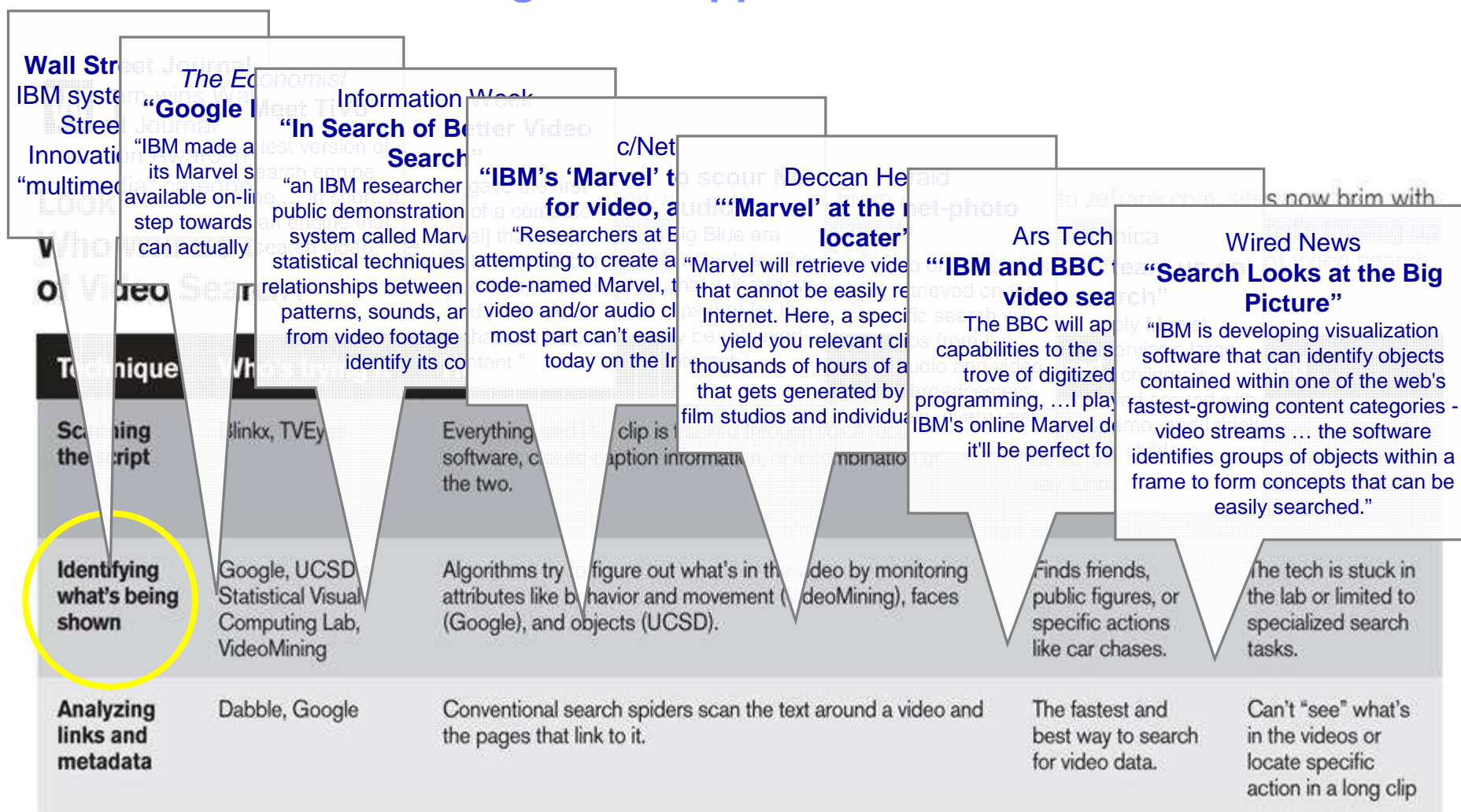
Deep archives, raw footage, independents

Consumer content

"Long tail"

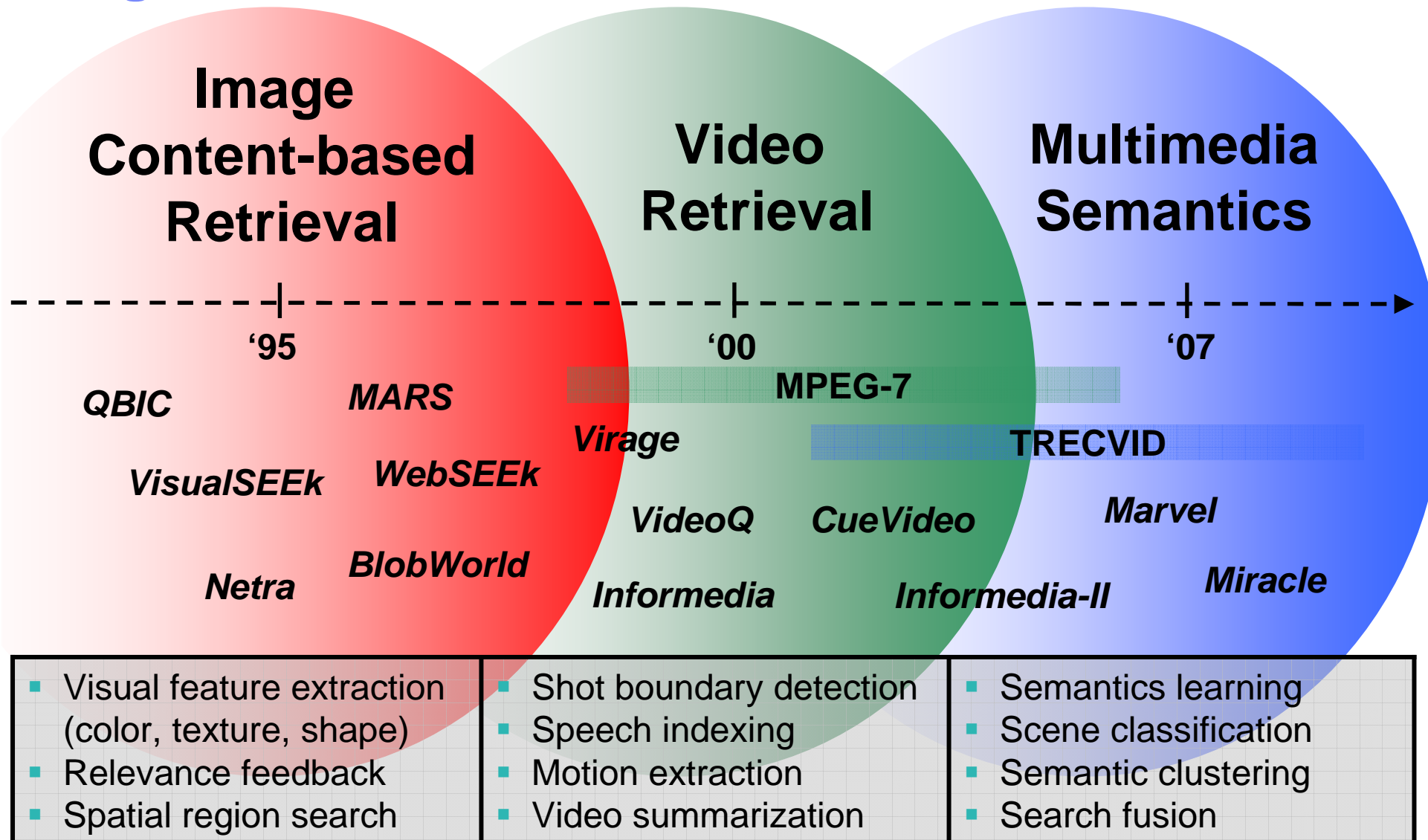
Digital item

The market is seeking a new approach for effective video search



IBM Multimedia Analysis and Retrieval System is recognized as leader in research and development of break-through techniques for video content-based analysis and search

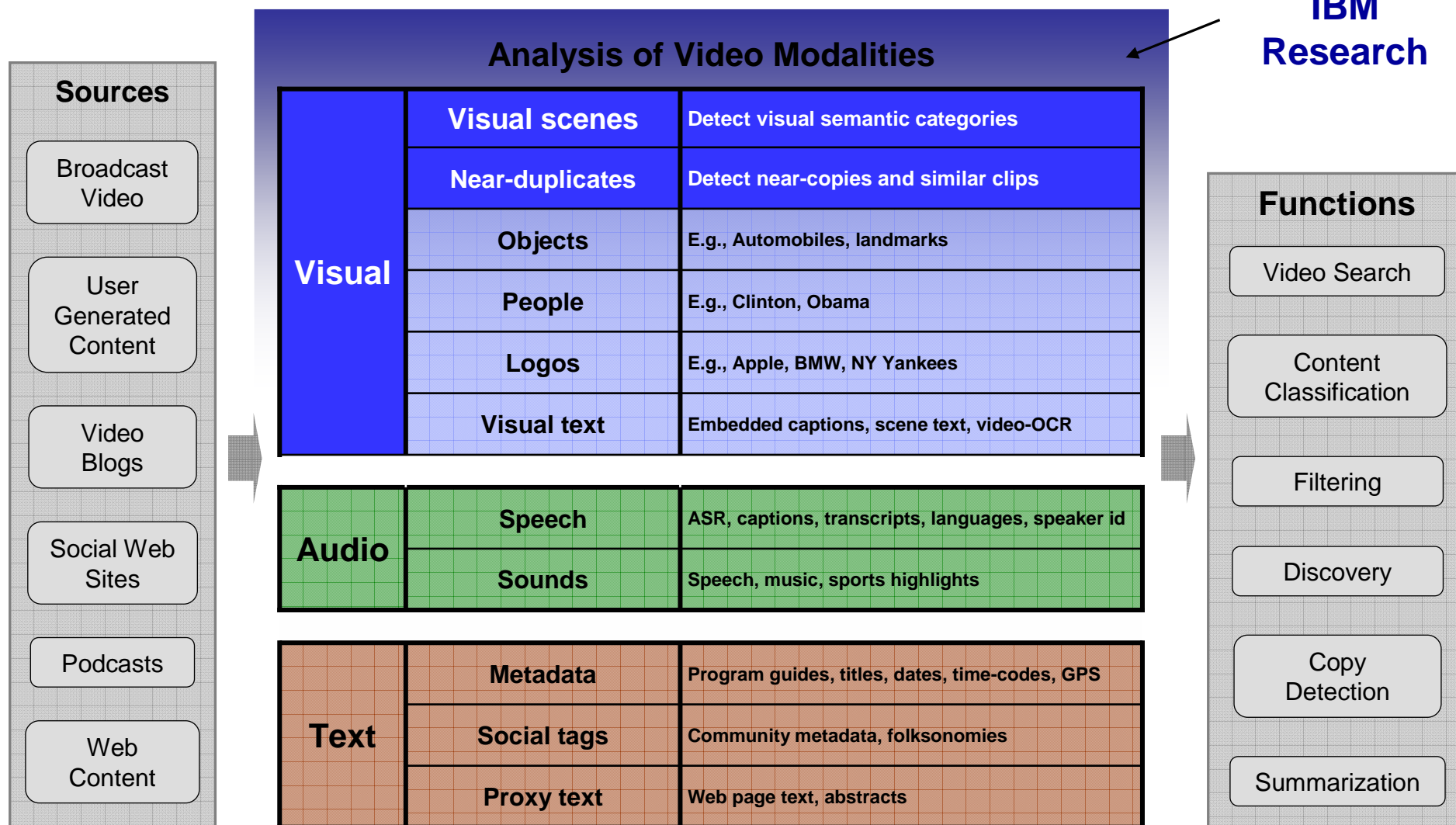
Progress in Multimedia Content-Based Retrieval



Bridging the Semantic Gap:
Analyze visual features and apply machine
learning techniques to classify video scenes
automatically

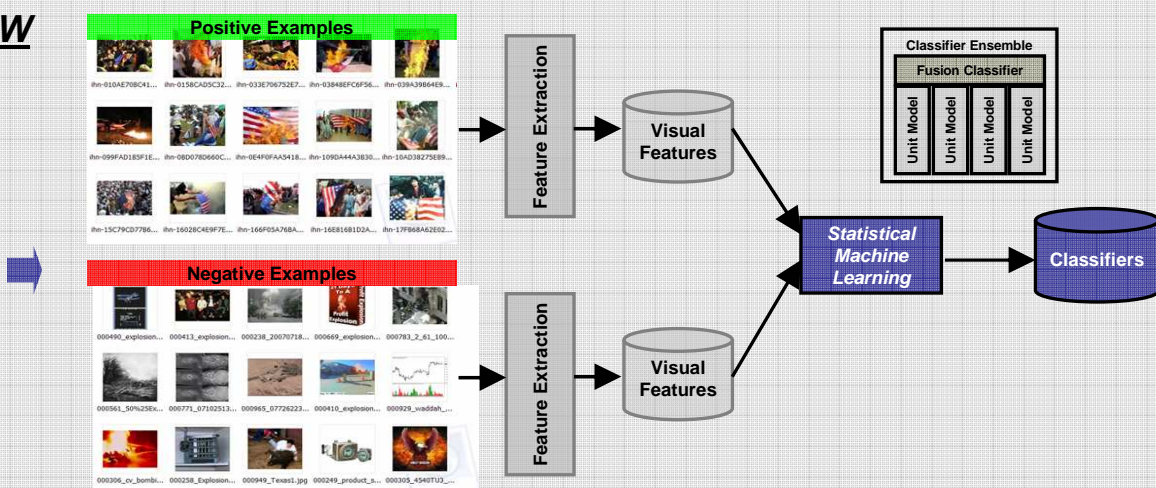
Making sense of the digital video chaos requires extracting meaningful information across multiple modalities (visual, audio, text, speech)

IBM
Research



Marvel – Software for learning visual categories and classifying and recognizing image/video content

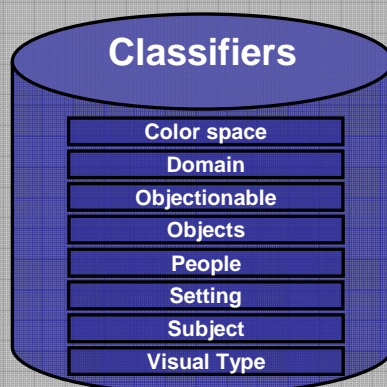
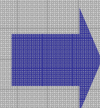
Image/Video Content Learning SW



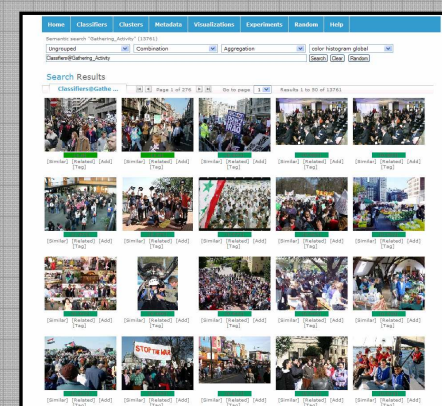
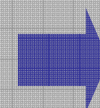
Image/Video Classification Software



Unknown
images/video



Taxonomy



Classified
images/video

Semantic models are created from training examples that are managed using multimedia taxonomies

MediaNet Knowledge Center

Scenes/Outdoors/Nature/Water/Waterviews

1 2 3 4

Add image

				
Delete	Delete	Delete	Delete	Delete
				
Delete	Delete	Delete	Delete	Delete
				
Delete	Delete	Delete	Delete	Delete

- Animals
- Art
- Entertainment
- Events
- Objects
- People
 - Children
 - Couple
 - Crowd
 - Face
 - Group
 - Person
 - Related
 - Women
- Places
- Scenes
 - Indoors
 - ManMadeSettings
 - Outdoors
 - Nature
 - Landscape
 - Mountain
 - Sky
 - Snow
 - Underwater
 - Vegetation
 - Water
 - Seasons
 - Urban
 - AncientStructures
 - Bridges
 - Buildings
 - Cityscape
 - Construction
 - ParkScenes
 - StreetScenes
 - WaterViews
 - Related
 - Scenes/Outdoors/Nature
 - Related
- Science
- Sports
- Types

Semantic Tagging of Multimedia Content

Video:



Associated speech:

Today, **jet fighters** practiced maneuvers and forces increased **military** preparations as tensions in **Middle East** reached ...

–Speech
–Closed Captions
–Transcript

Text Analysis
(optional)

“Jet fighters”

“military”

“Middle East”

Text Metadata

Image & Video
Sequence Analysis

Visual Features

Motion

Color patterns

Texture

Shapes

Visual Feature Metadata

Automatic Semantic Concept Detection

Semantic
Models

Airplane (0.8)

Outdoors (0.9)

People (0.7)

Protest (0.6)

Sky (0.7)

Parade (0.5)

Meeting (0.7)

Indoors (0.8)

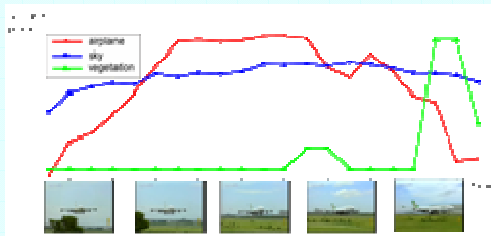
Semantics
Metadata

* with associated
confidence scores

Multimedia Semantic Analysis and Search

Semantics Modeling:

- Modeling Large-scale Semantic Spaces (1)
- Multi-Granular & Parts-based Modeling
- Temporal Event Semantic Modeling
- Sequence Rhythm



Content Extraction:

- (2) • Semantic Labeling of Visual Clusters
- Temporal Pattern Mining
- Cross-Channel Topic Tracking



Nature, Day, Outdoors

User

Model

Extract

Search

Searching, Browsing & Interaction:

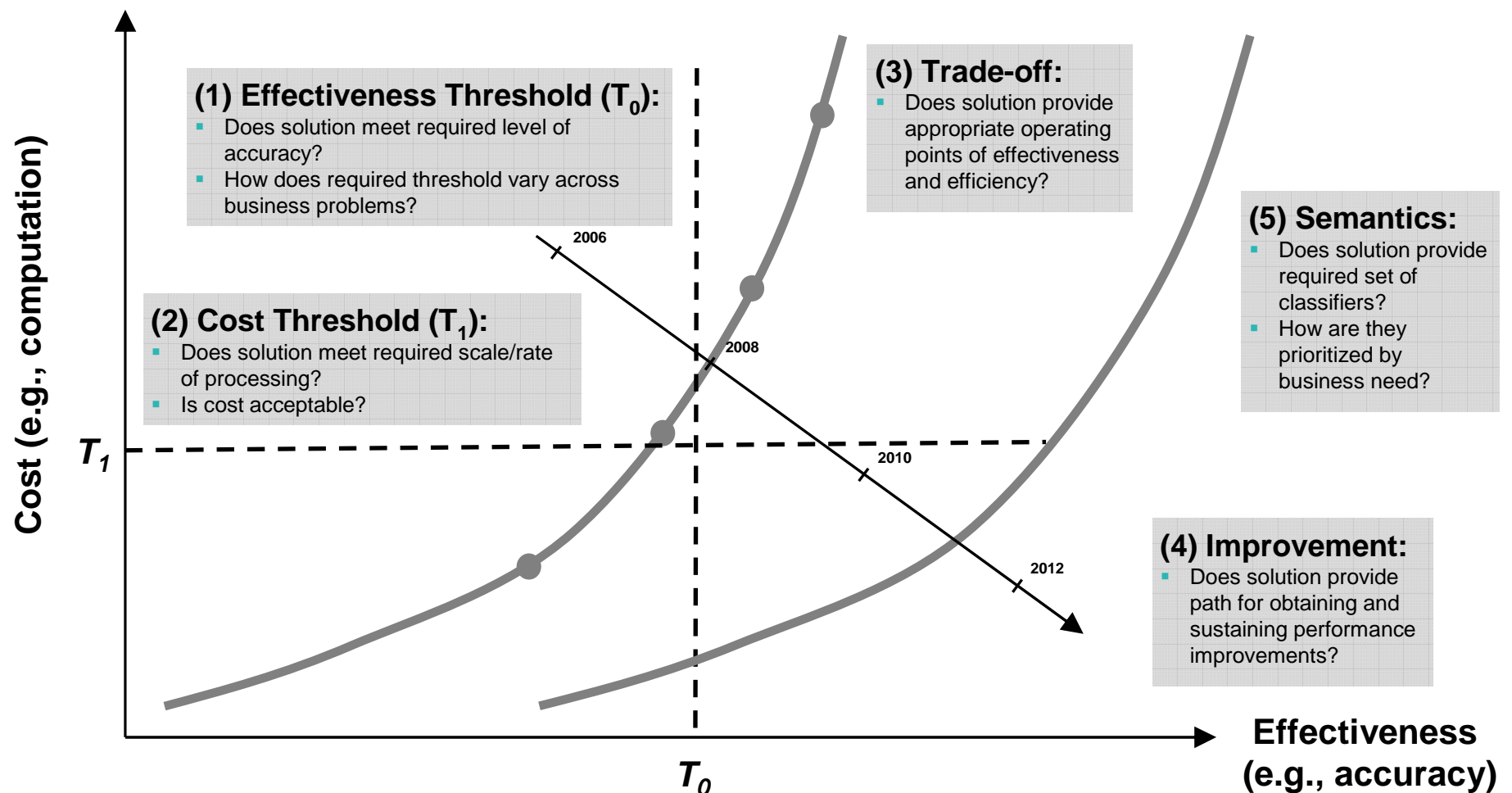
- Multi-Modal Search and Retrieval (3)
- Query Expansion for Multimodal Video Retrieval (4)
- Query-Class Dependent Video Search



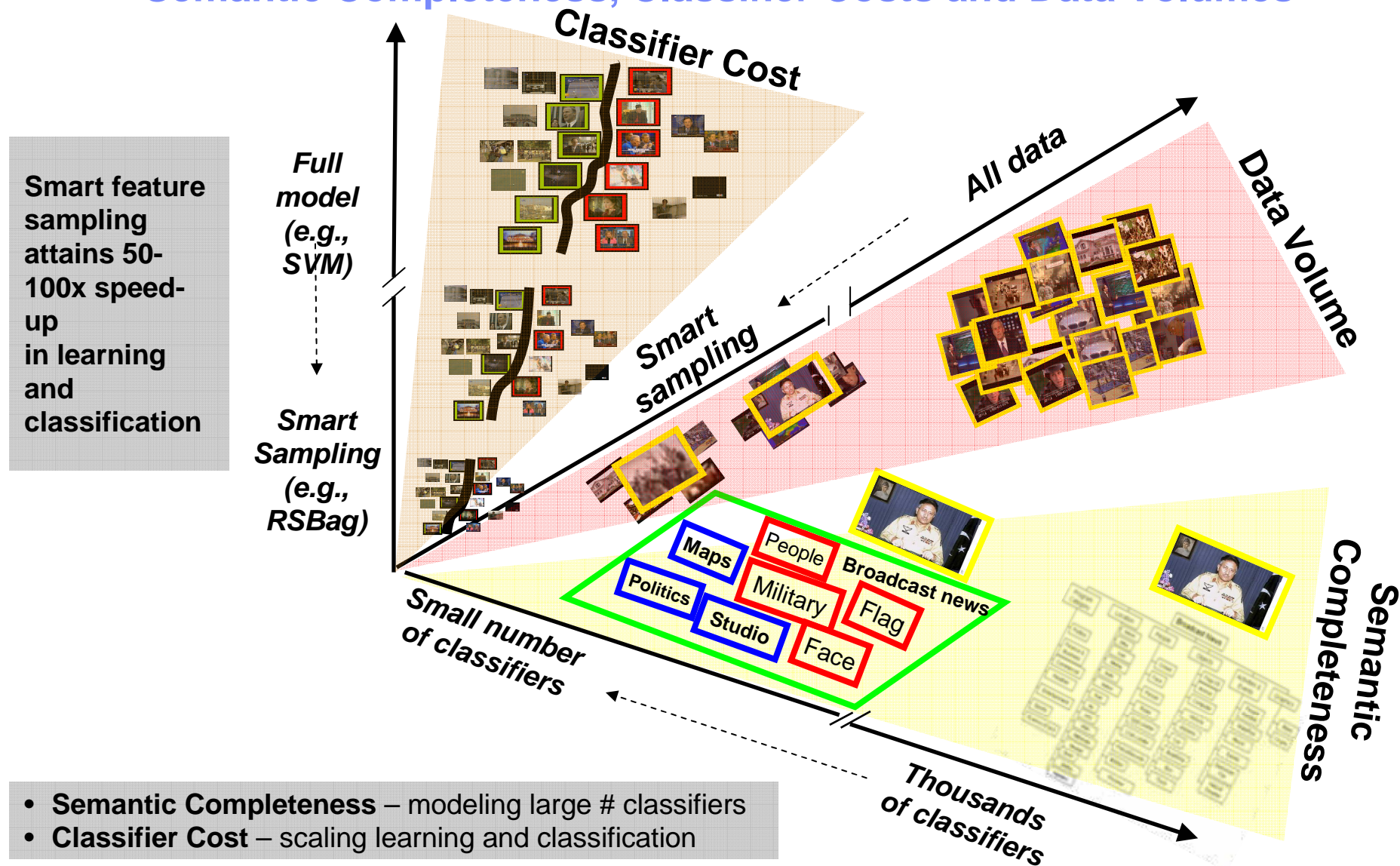
(1) Modeling Large-scale Semantic Spaces

Challenge with any nascent technology is to fit to suitable problem set

■ Five Dimensions of Visual Recognition Performance



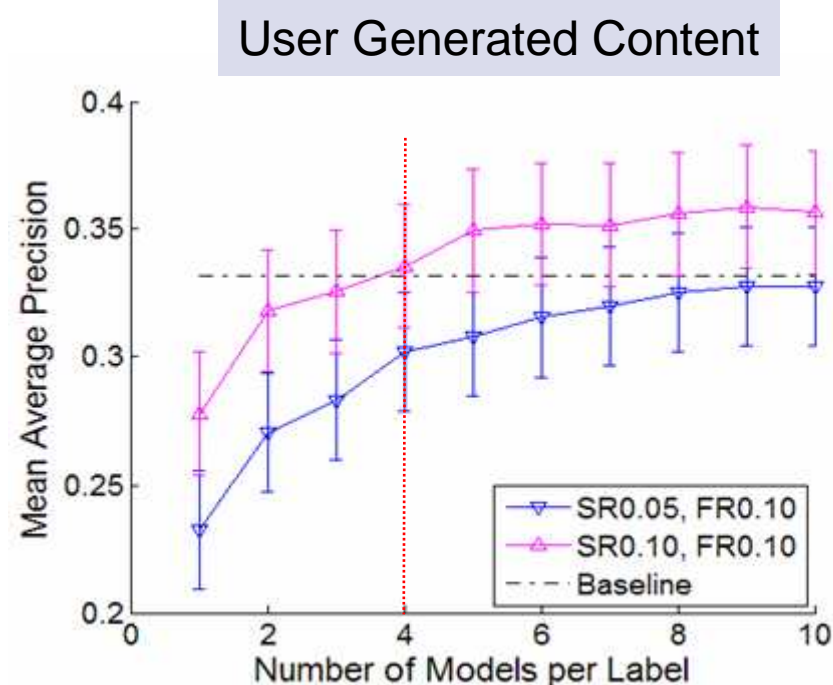
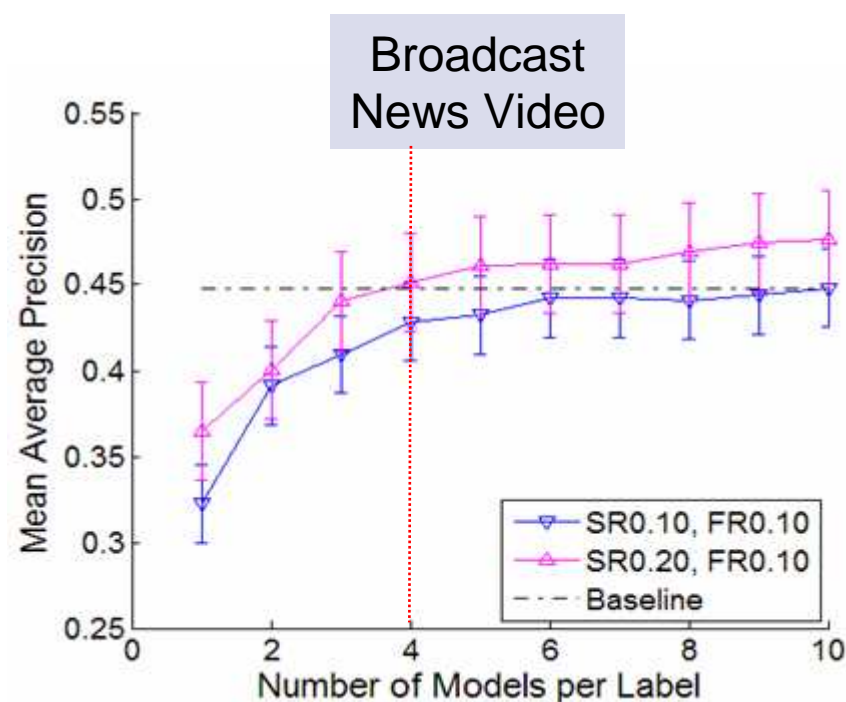
Scalability in Visual Semantic Classification is Achieved by Trading-off Semantic Completeness, Classifier Costs and Data Volumes



Smart feature sampling during learning of visual semantic classifiers allows efficient scaling to large number of video semantic classifiers

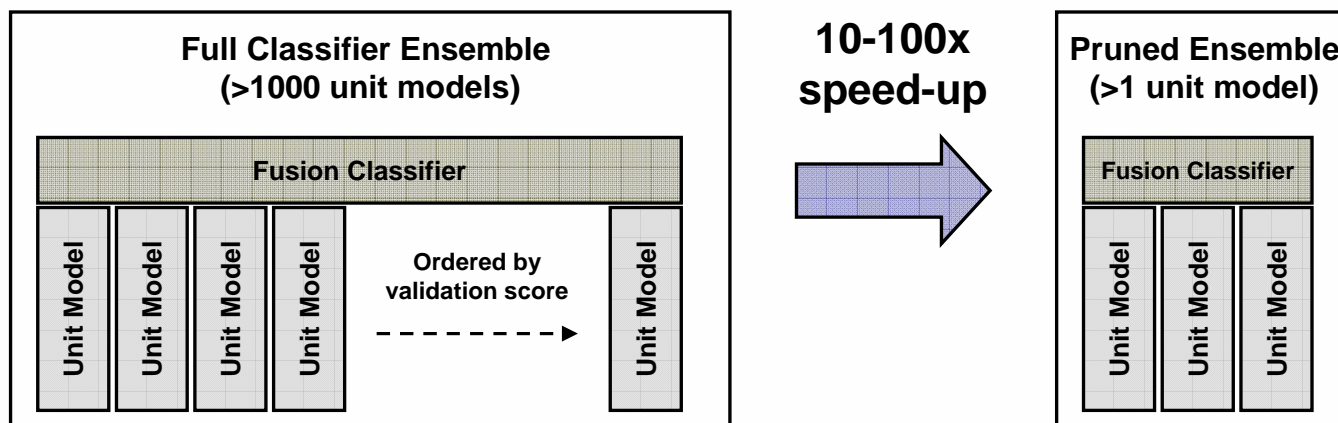
- Smart feature sampling of features greatly speeds-up learning and classification
- Easy-to-use trade-off of classification accuracy and computation
- Unit models can be leveraged across multiple semantic concepts for greater efficiency

*** Classification Accuracy reaches high value using small number of unit models**

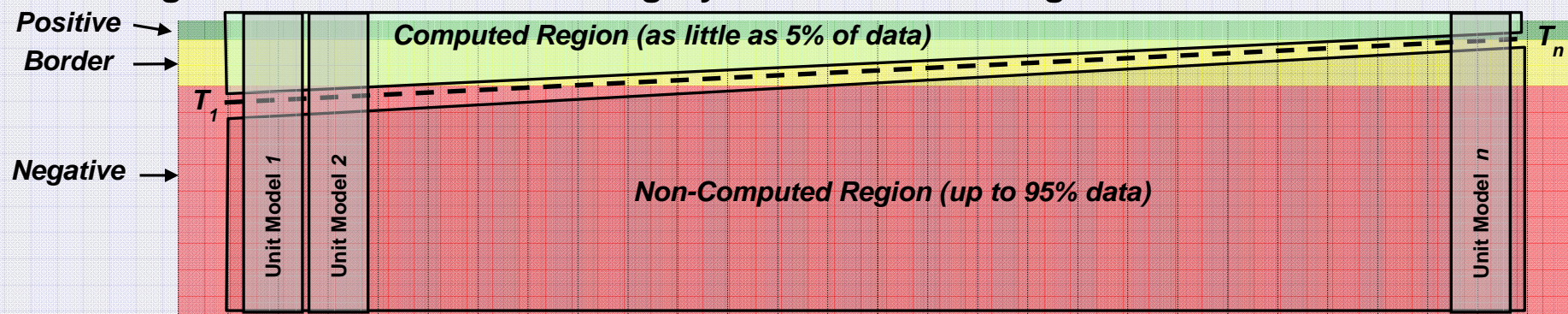


IBM's solution uses a highly granular ensemble classifier approach built on 140 visual descriptors that supports large-scale processing through progressive classification and run-time trade-off in accuracy and speed

Classifier Trade-off (Speed vs. Accuracy)



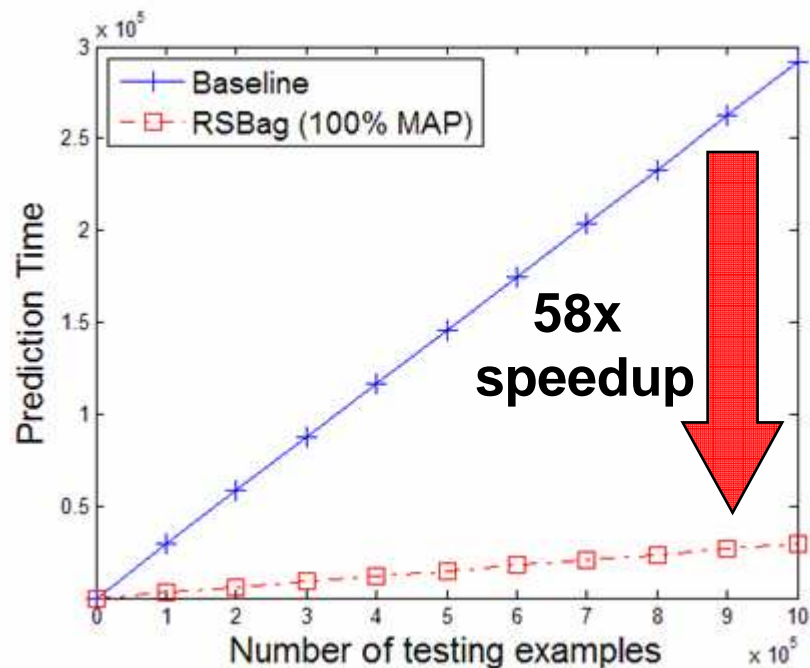
Progressive Classification using Dynamic Thresholding in Classifier Ensemble



Significant speed-up in learning makes it possible to learn new visual semantic models in near real-time as needed

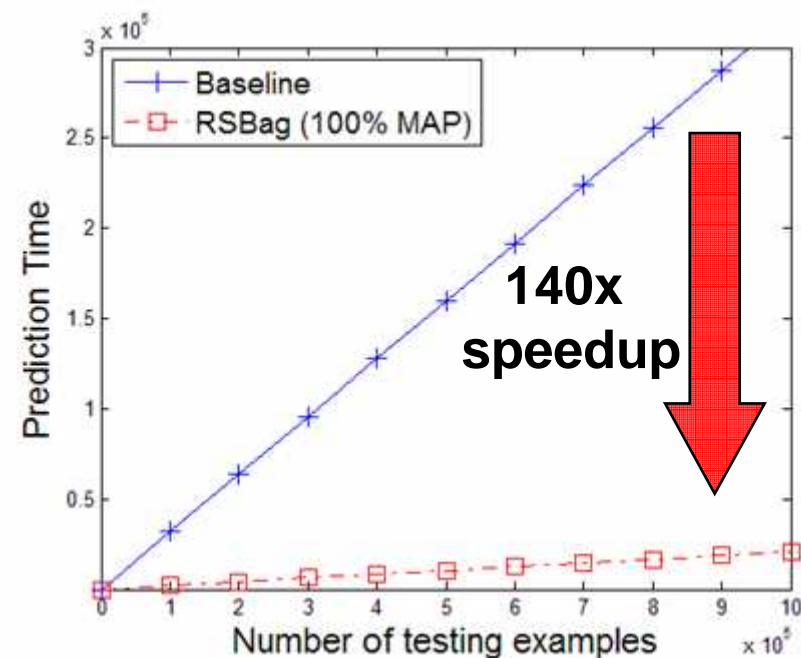
Broadcast News Video

Training Time	All data	Smart Sampling
	98 min	101 sec



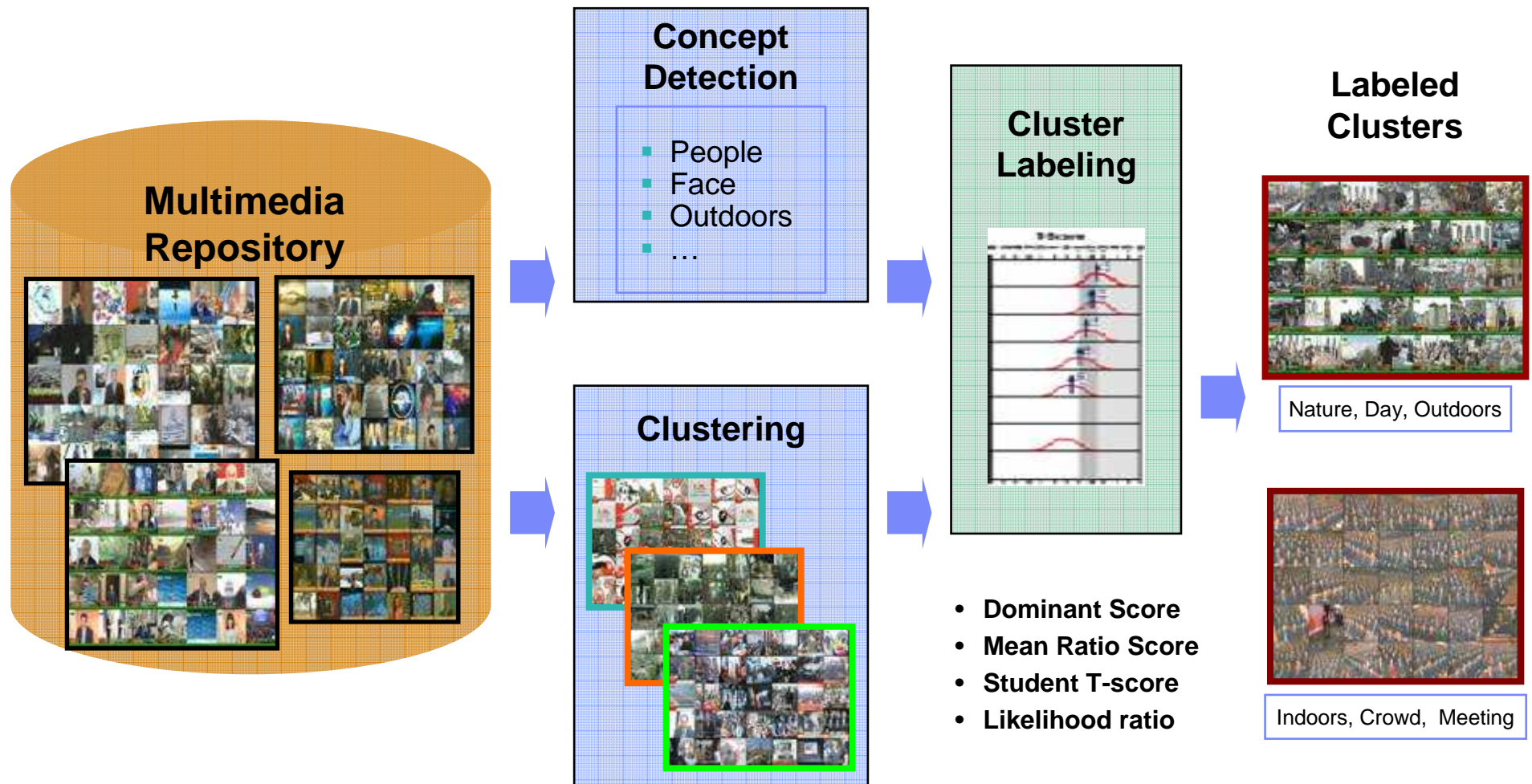
User Generated Content

Training Time	All data	Smart Sampling
	85 min	36 sec



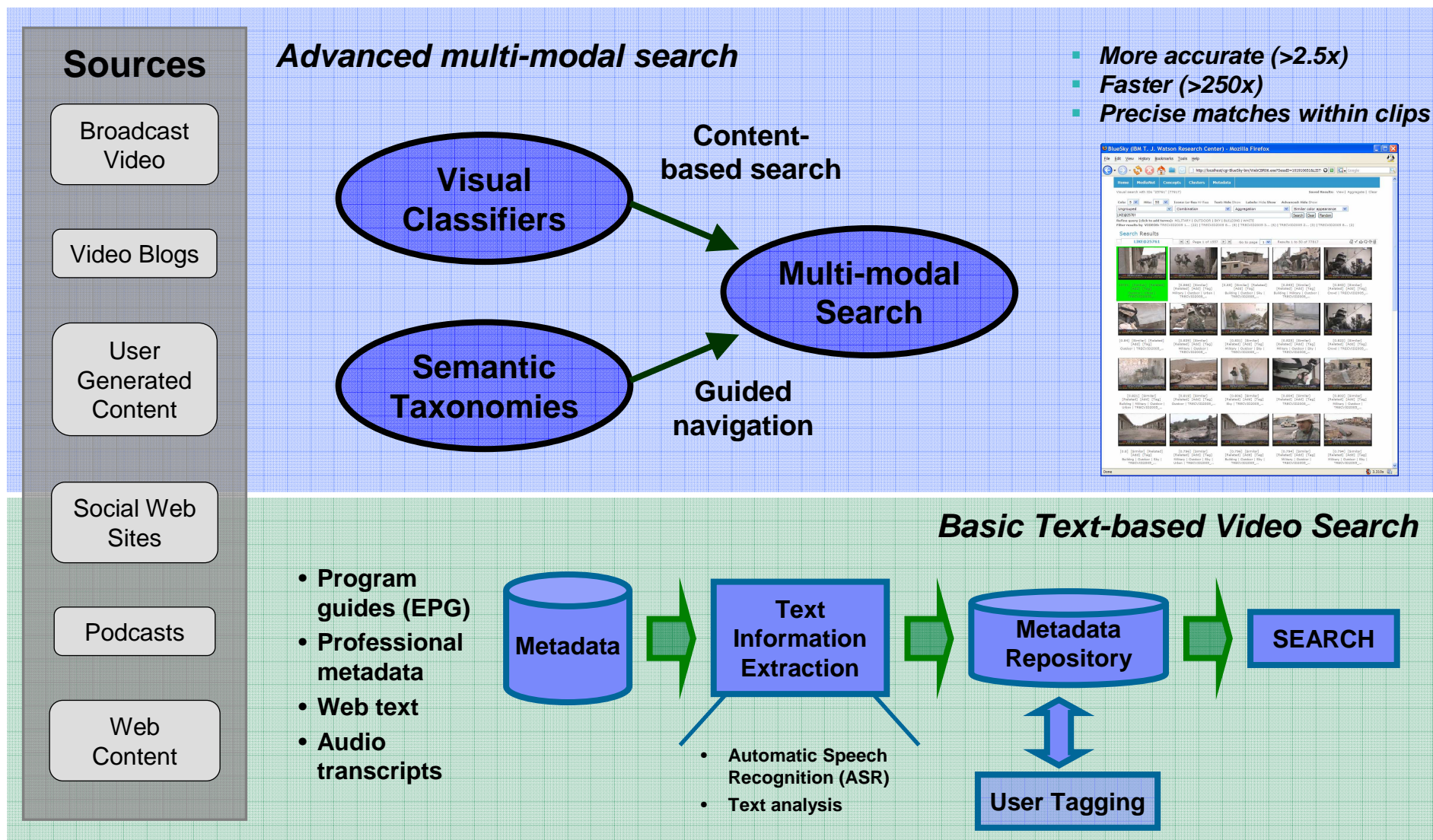
(2) Semantic Labeling of Visual Clusters

Semantic Labeling of Visual Clusters – Discovering Descriptive & Discriminative Semantics (*ICME-2006*)



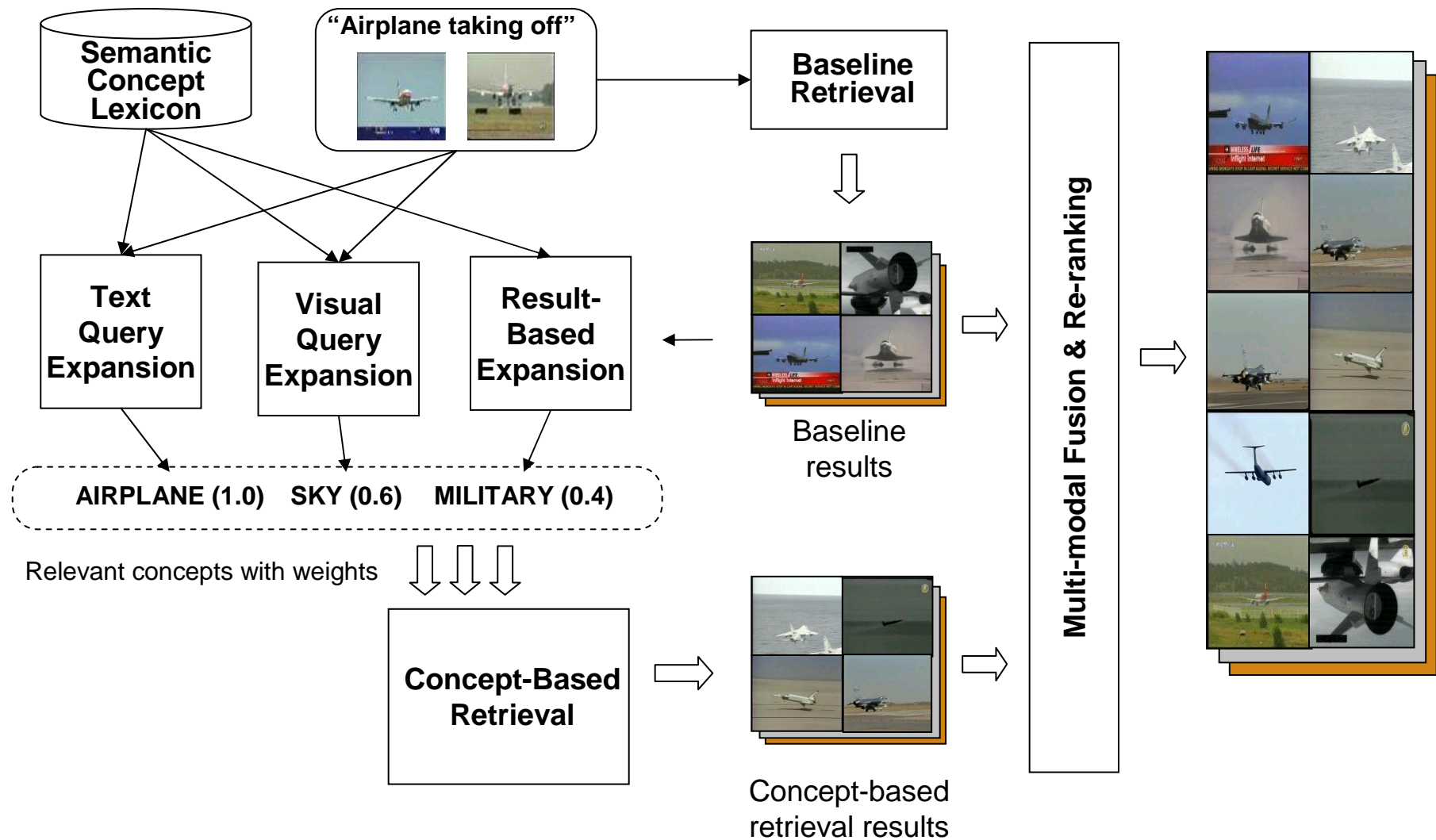
(3) Multi-Modal Search and Retrieval

IBM's "content-based" approach improves video analysis by classifying scenes visually and allows multi-modal search of video content

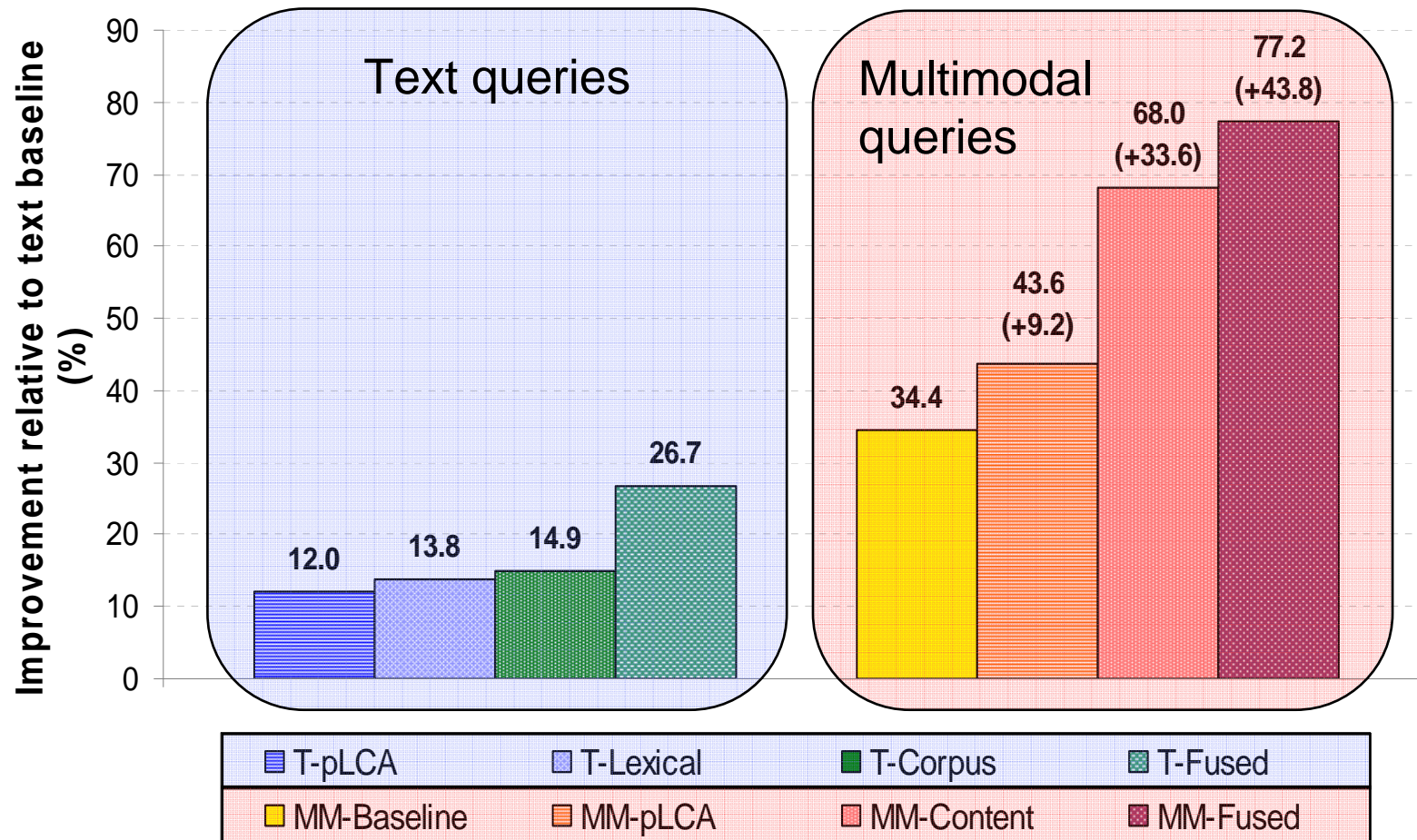


(4) Query Expansion for Multi-modal Video Retrieval

Query Expansion for Multi-modal Video Retrieval (*ACM Multimedia, Sept. 2007*)

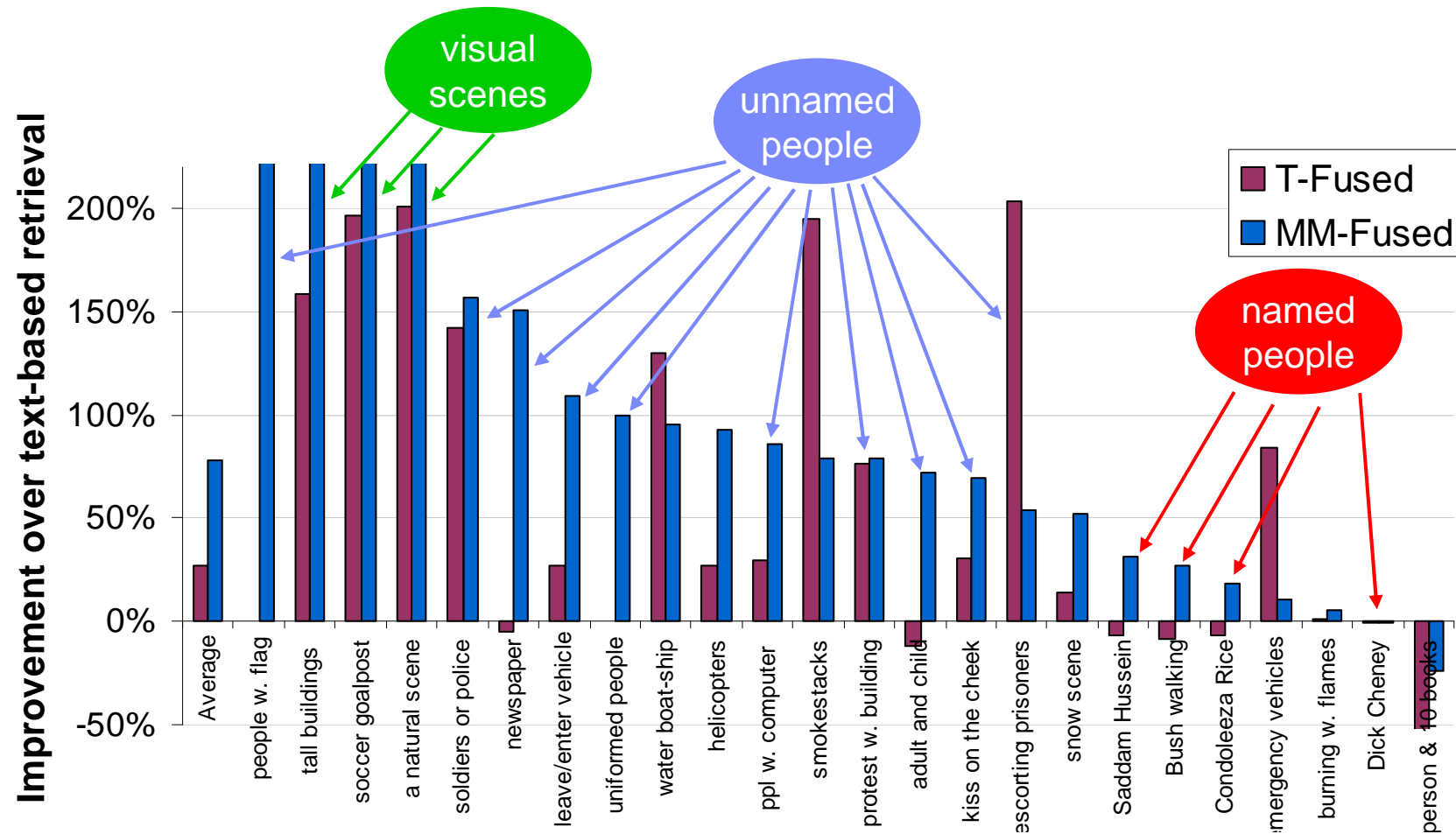


Empirical Evaluation & Comparison (TRECVID 2006 data)



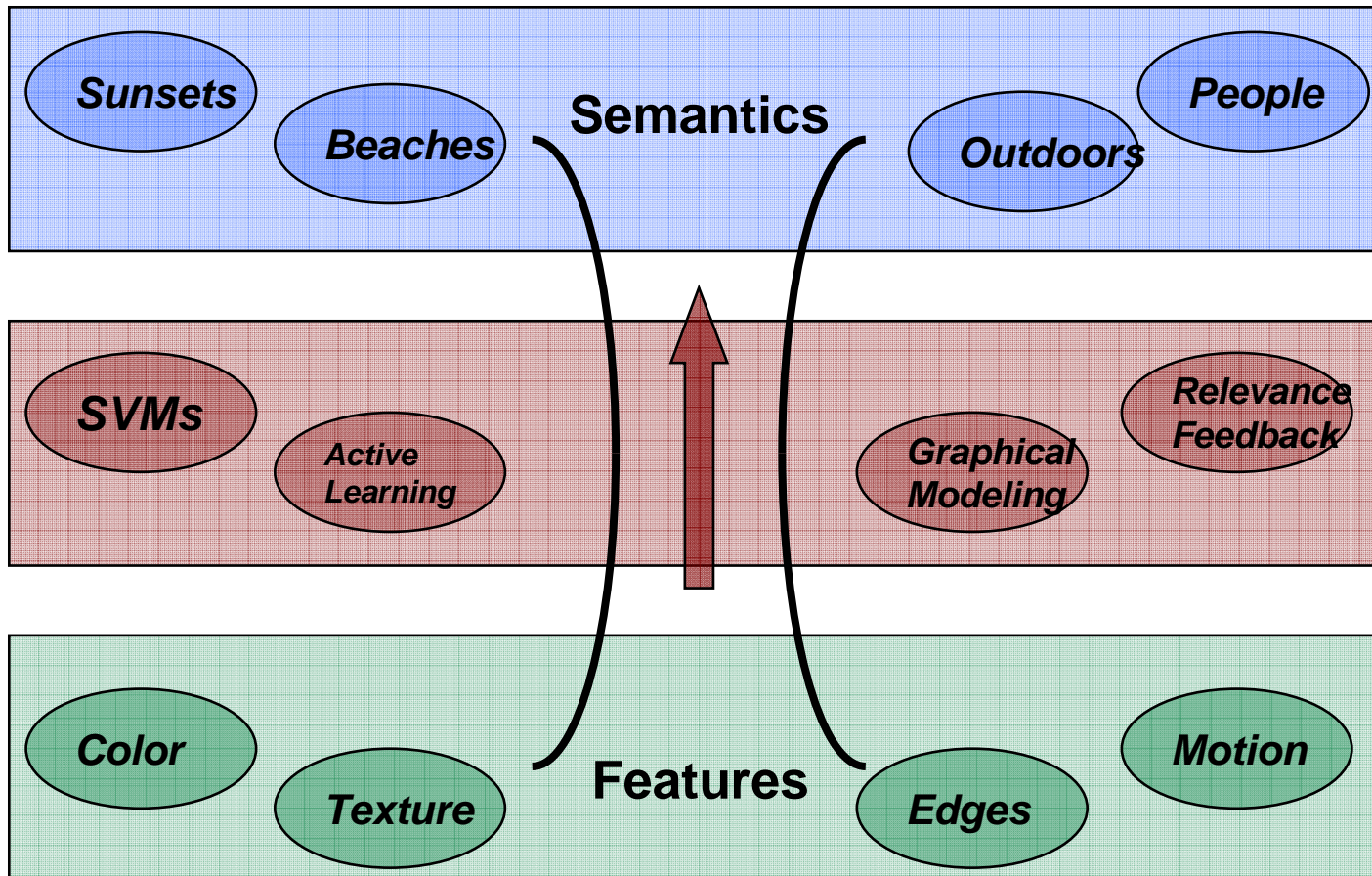
- Text-based expansion approaches perform comparably but are complementary
- Content-based approaches bring significant further improvements

Empirical Evaluation & Comparison (Cont'd)



Related efforts on the modeling of large video semantic spaces

Bridging the Multimedia Semantic Gap – What's the Destination?

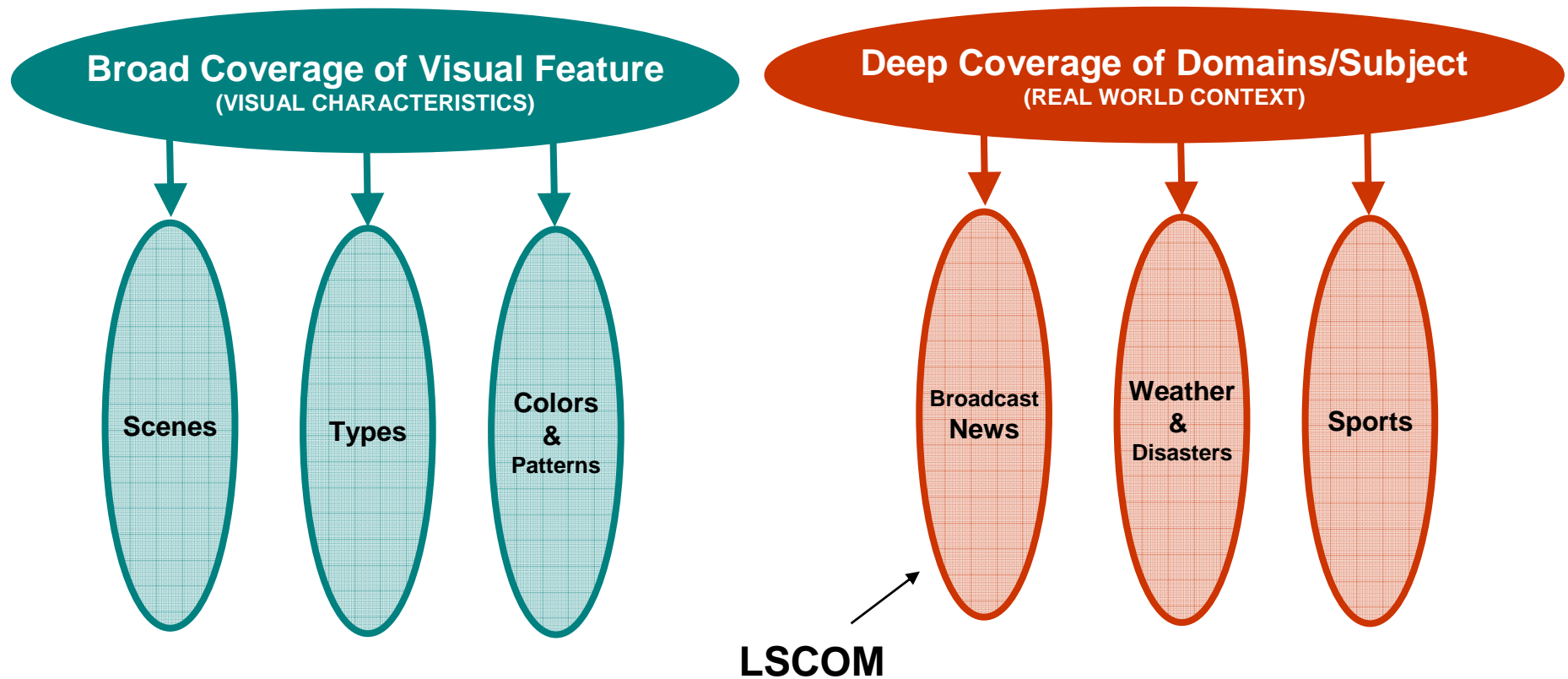


- How do we fully develop the semantic space itself?
- Research is producing powerful learning tools
- Foundation established (e.g., MPEG-7)

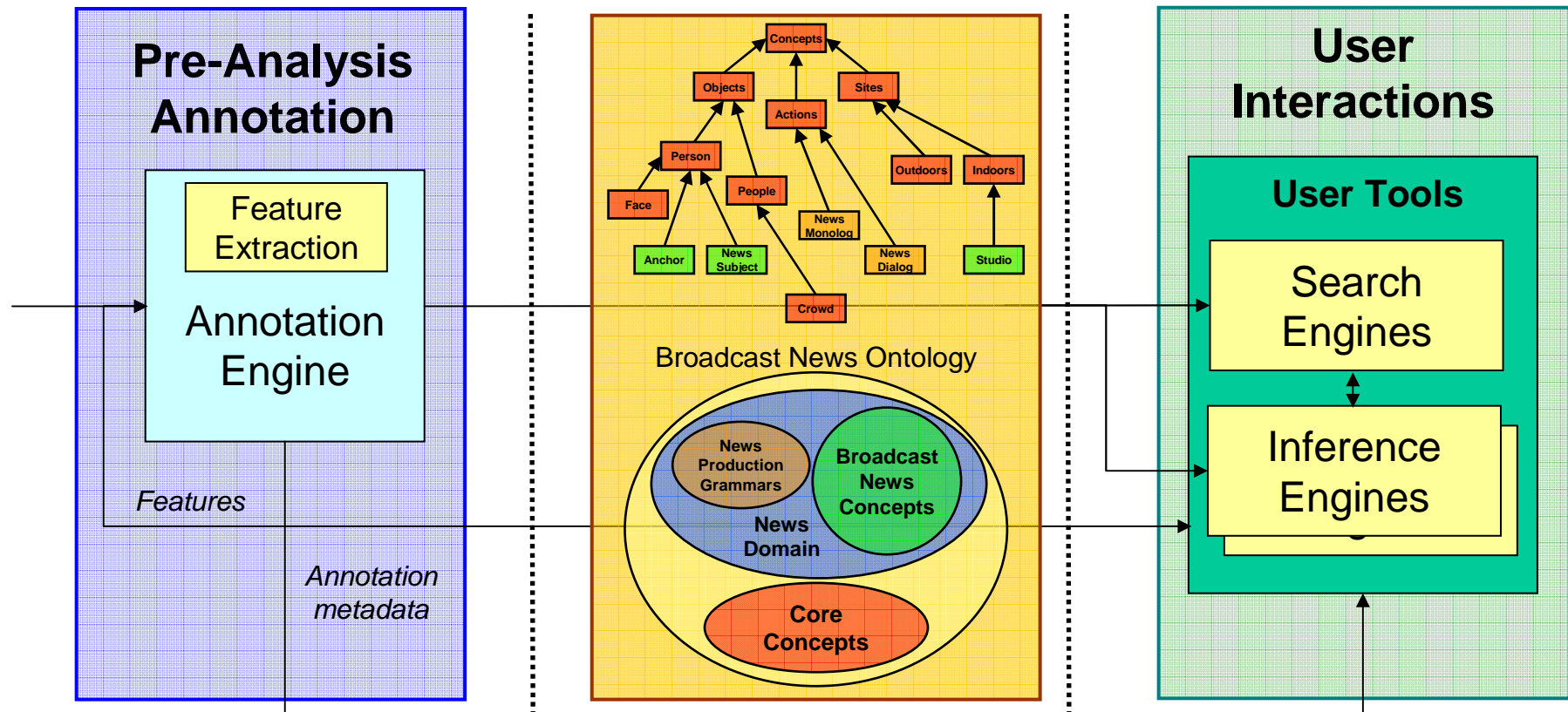
- Working on the foundation and the bridge, but what is the ultimate destination?
- Don't want to build a bridge to nowhere !!!

Structuring Multimedia Semantic Spaces

- Multimedia ontologies resemble faceted taxonomies but use richer semantic relationships among nodes that contain multimedia signifiers
- Can be developed to support different perspectives on multimedia content (i.e. visual characteristics vs. subject hierarchy)



Large Scale Concept Ontology for Multimedia Understanding (LSCOM*) – 1,000 Semantic Concepts



- **LSCOM** is collaborative effort to develop a large standardized taxonomy for describing multimedia broadcast news video
- **Designed to optimize:** (1) utility for facilitating end-user access, (2) coverage of large semantic space, (3) feasibility for automated extraction, (4) observability in diverse multimedia broadcast news data sets

5/8/2008

IBM Research | IBM Confidential

37

* M. Naphade, J.R. Smith, J. Tesic, S.-F. Chang, W. Hsu, L. Kennedy, A. Hauptmann, J. Curtis, *IEEE MultiMedia*, July 2006


Large Scale Concept Ontology for Multimedia (LSCOM)

LSCOM Lexicon Definitions and Annotations

DTO Challenge Workshop on Large Scale Concept Ontology for Multimedia

Quick Guide to LSCOM Data Sets

1. **LSCOM: a collection of annotations for 449 concepts.**

 [Download the LSCOM annotation data.](#) (103 MB file. Expands to 2.54 GB on disk.)

LSCOM Citation: LSCOM Lexicon Definitions and Annotations Version 1.0, DTO Challenge Workshop on Large Scale Concept Ontology for Multimedia, Columbia University ADVENT Technical Report #217-2006-3, March 2006. [\[pdf\]](#)

Lexicon information. (Detailed list of annotated concepts.)

2. **LSCOM-Lite: an overlapping precursor to LSCOM with annotations for 39 concepts.**

 [Download the LSCOM-lite annotation data.](#) (16 MB file.)

LSCOM-Lite Citation: M. R. Naphade, L. Kennedy, J. R. Kender, S.-F. Chang, J. R. Smith, P. Over, and A. Hauptmann, "A Light Scale Concept Ontology for Multimedia Understanding for TRECVID 2005," IBM Research Technical Report, 2005. [\[pdf\]](#)

3. **LSCOM Revised Event/Activity Annotations: video-based re-labeling of 24 LSCOM concepts.**

 [Download the LSCOM Revised Event/Activity annotations.](#) (236 KB file.)

LSCOM Revised Event/Activity Annotations Citation: Lyndon Kennedy, Revision of LSCOM Event/Activity Annotations, DTO Challenge Workshop on Large Scale Concept Ontology for Multimedia, Columbia University ADVENT Technical Report #221-2006-7, December 2006. [\[pdf\]](#)

Summary

The DTO sponsored LSCOM workshop has developed an expanded multimedia concept lexicon on the order of 1000. Concepts related to events, objects, locations, people, and programs have been selected following a multi-step process involving input solicitation, expert critiquing, comparison with related ontologies, and performance evaluation. Participants of the process include representatives from intelligence community users, ontology specialists, and multimedia analytics researchers. In addition, each concept has been qualitatively assessed according to some criteria, such as utility (usefulness), observability (by humans), and feasibility (by automatic detection). An annotation process was completed in late 2005 by student annotators at Columbia University and CMU, over the entire development set of TRECVID 2005 videos. Human subjects judge the presence or absence of each concept in the key frame of each subshot, resulting in a total of 61901 labels for each concept.

The first version of the **LSCOM annotations** [3] consist of keyframe-based labels for 449 visual concepts, out of the 834 initial selected concepts, over the entire TRECVID 2005 development set (61901 subshots).

- **What is it?** – lexicon covering large semantic space for broadcast news analysis from IC perspective
 - >1,000 concepts
 - Large annotated video data set (449 visual concepts, 24 temporal activities)
- **Impact to-date:**
 - LSCOM-lite used in TRECVID
 - Downloaded by >170 groups
- **Available for download:**
 - LSCOM lexicon
 - LSCOM annotations
 - “Columbia374” SVM models

www.ee.columbia.edu/dvmm/lscm

Sample of 170+ institutions downloading LSCOM

- Yahoo! Research
- Intel
- AT&T
- FXPAL
- University of Amsterdam
- Oxford University
- Nanyang Technological University, Singapore
- National Taiwan University
- Tsinghua University
- KDDI, Japan
- Dublin City University, Ireland
- University of Central Florida
- University of Texas, Austin
- UC Berkeley
- Others ...

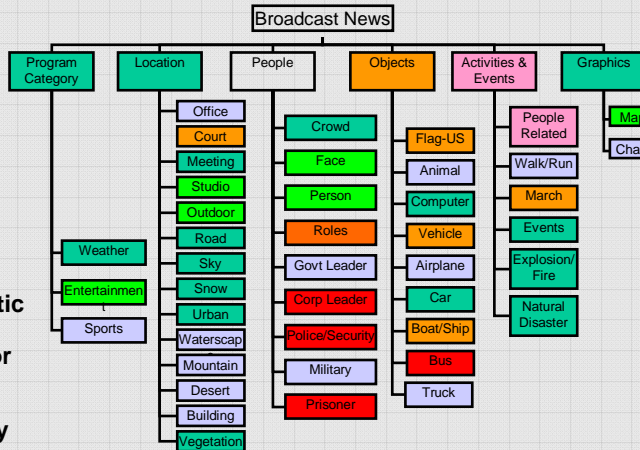
[Link to download log](#)

Large Scale Concept Ontology for Multimedia Understanding (LSCOM*) – 1,000 Semantic Concepts

Design

Taxonomy Design:

- *LSCOM 1,000 Semantics Concepts
- Designed to optimize:
 1. Utility for facilitating user access
 2. Coverage of large semantic space
 3. Feasibility for automated extraction
 4. Observability in diverse broadcast news data sets

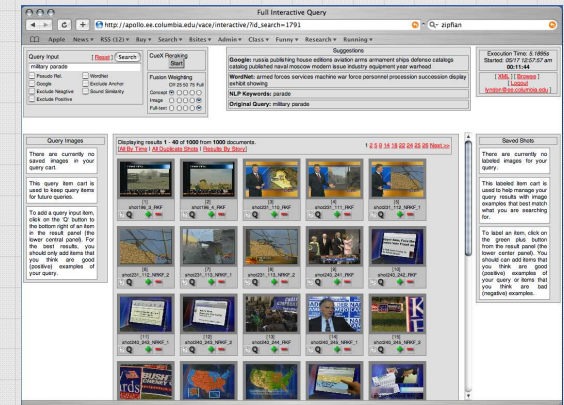


* IEEE MultiMedia, Summer 2006

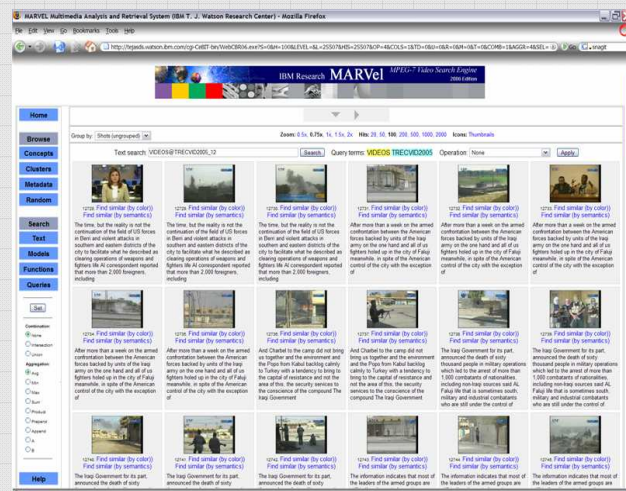
Annotated Concepts:

- Event/Activity (56 - 13%) - *Airplane taking off, car crash, shaking hands*
- People (113 - 25%) - *Female person, firefighter, judge*
- Location (89 - 20%) - *Hospital, airfield, cityscape*
- Object (135 - 30%) - *Power plant, tent, vehicle*
- Scene (49 - 10%) - *Vegetation, interview, urban*
- Program (7 - 2%) - *Entertainment, weather, finance*

Annotate



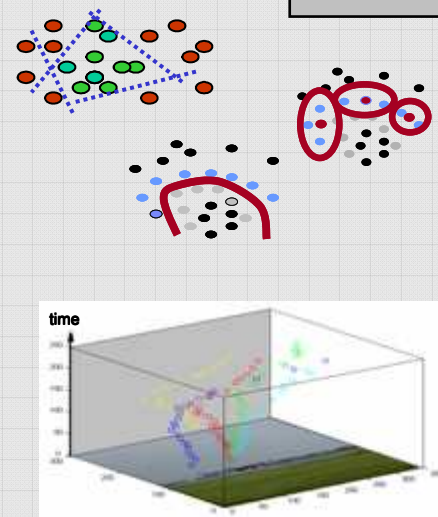
Use



Search & Retrieval:

- Use-case driven assessment for searching & topic threading
- Multiple search engine implementation and evaluation
- Support for automatic, manual and interactive search

Model



Semantics Modeling:

- Appearance-based feature extraction (color, texture, shape, edges, motion)
- Machine learning and statistical modeling (SVMs, GMMs, Nearest Neighbor)
- Multi-feature and multi-model fusion
- Scalable modeling using a massive distributed computing infrastructure

Public evaluations such as TRECVID

NIST TRECVID Video Retrieval Benchmark at a Glance

- **TRECVID:**
 - NIST benchmark for evaluating state of the art in video retrieval
- **Benchmark tasks:**
 - Shot Boundary Determination
 - Semantic Concept Detection
 - Story Segmentation
 - Search



Topic 101: Find shots of a basket being made - the basketball passes down through the hoop and net

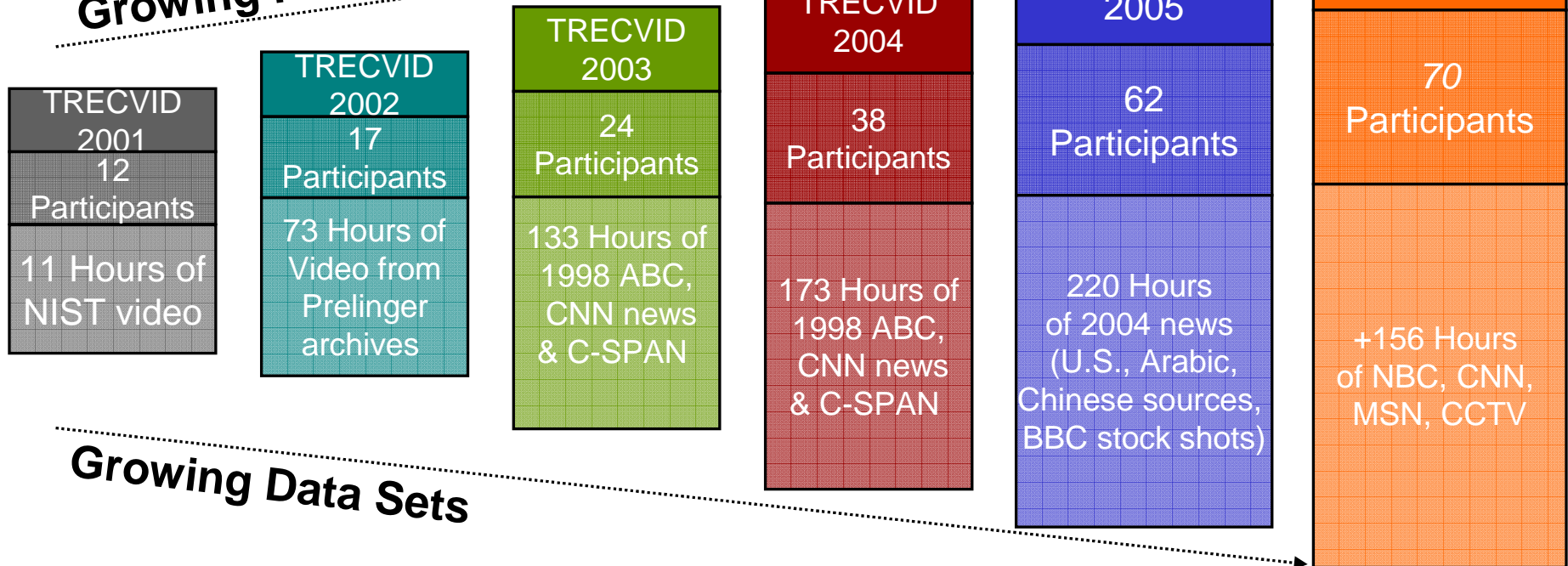


Topic 129: Find shots zooming in on the US Capitol dome.



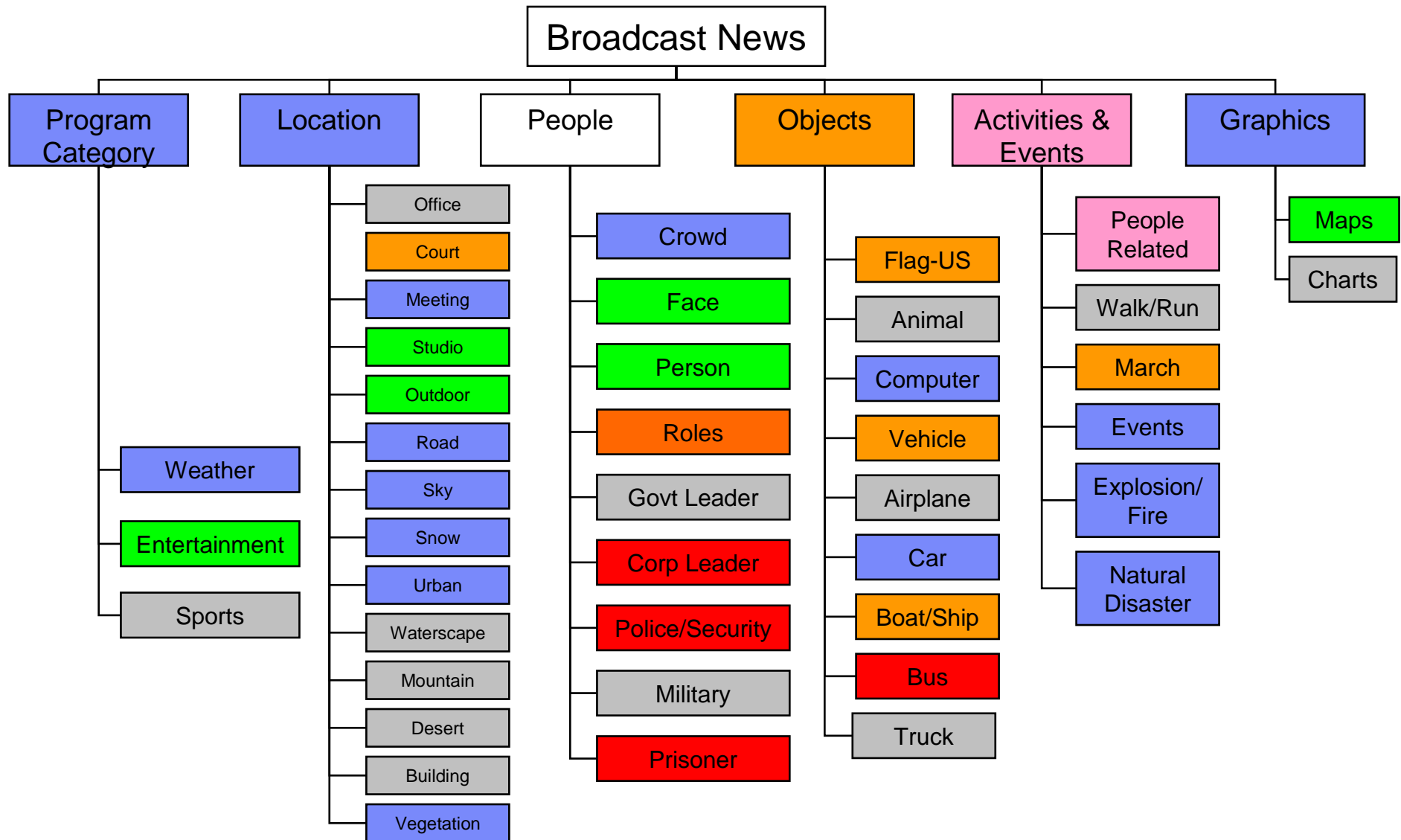
Topic 104 and 167: Find shots of an airplane taking off

Growing Participation



Growing Data Sets

“LSCOM-lite” for TRECVID High-Level Feature Detection



TRECVID

Video corpus

- Broadcast news from U.S., Arabic, and Chinese sources
 - TRECVID 2005: 160 hrs
 - TRECVID 2006: 240 hrs
- Speech transcripts based on
 - Speech Recognition
 - Machine Translation

Query topics

- Brief description of topic
- 5-10 visual examples/topic
- 24-25 topics each year
- Typical topic classes:
 - Named people (Person-X)
 - Generic people interactions
 - Sports
 - Objects/Events
 - Scenes/settings

Search types

- Automatic, manual, interactive



Topic 149:
Find shots of Condoleezza Rice



Topic 150: Find shots of Iyad Allawi, the former prime minister of Iraq



Topic 151: Find shots of Omar Karami, the former prime minister of Lebanon



Topic 152: Find shots of Hu Jintao, president of the People's Republic of China



Topic 153: Find shots of Tony Blair.



Topic 159: Find shots of George W. Bush entering or leaving a vehicle (e.g., car, van, airplane, helicopter, etc) (he and vehicle both visible at same time)



Topic 157: Find shots of people shaking hands



Topic 161: Find shots of people with banners or signs



Topic 163: Find shots of a meeting with a large table and more than two people



Topic 165: Find shots of basketball players on the court



Topic 171: Find shots of a goal being made in a soccer match



Topic 156: Find shots of tennis players on the court both players visible at same time



Topic 158: Find shots of a helicopter in flight



Topic 164: Find shots of a ship or boat



Topic 167: Find shots of an airplane taking off



Topic 160: Find shots of something (e.g., vehicle, aircraft, building) on fire with flames & smoke visible



Topic 168: Find shots of a road with one or more cars



Topic 170: Find shots of a tall building (with more than 5 floors above the ground)

ACM CIVR'07 VideOlympics Showcase (July, 2007)

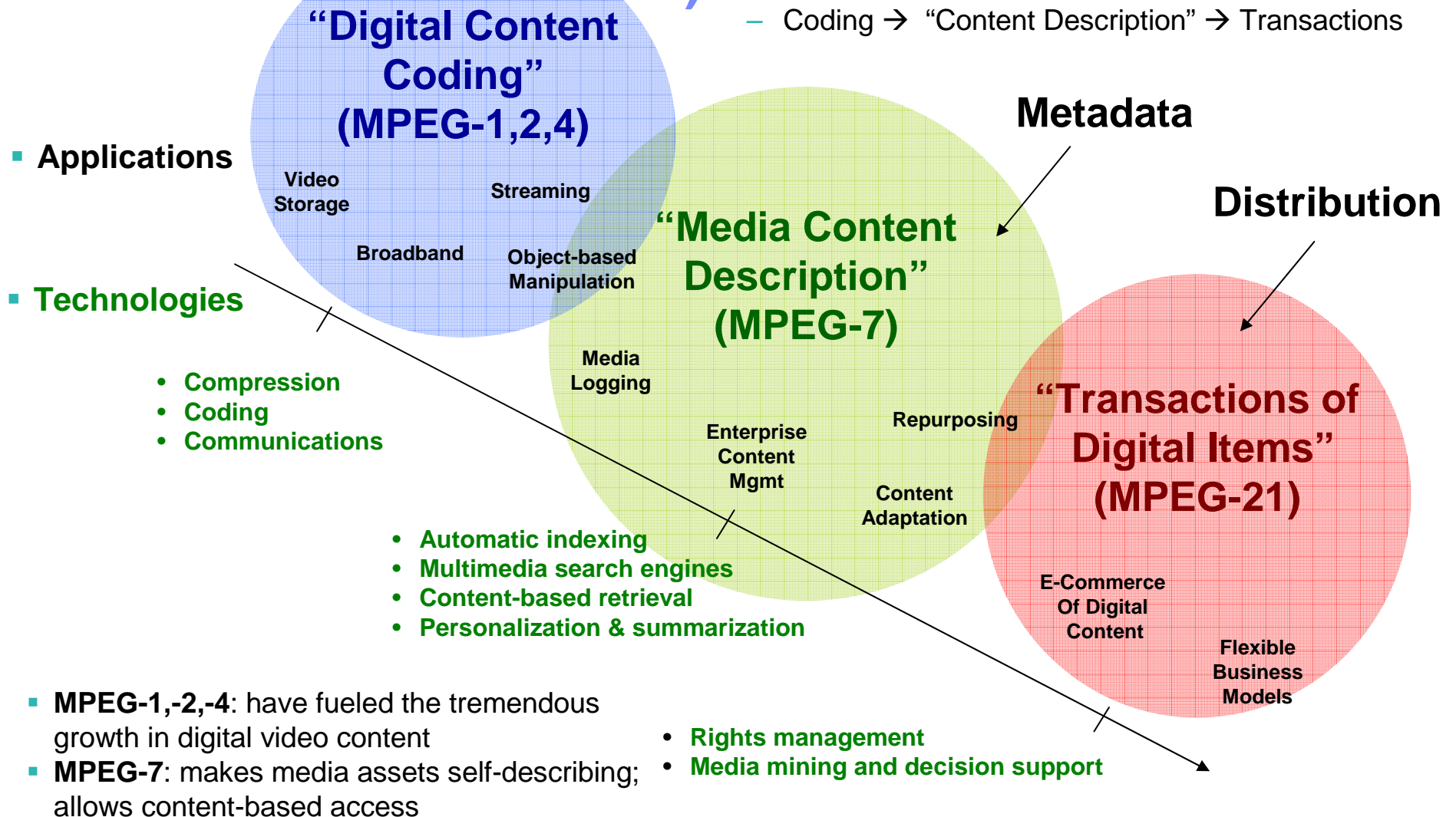
- A video search showcase that goes beyond the regular demo session and a small size of TRECVID participants
 - The showcase participants will simultaneously do an **interactive search task** during the VideOlympics showcase event.
 - Paul Over from NIST will provide **text-only** search topics onsite
 - Unlike TRECVID, results are submitted immediately after they are found.
 - Fun to do for the participants and fun to watch for the conference audience
- The first VideOlympics event is a great success
 - 9 retrieval systems submitted from worldwide participants and great interest from the audience in the conference
- Video: <http://videolympics.org/>
- Next year: CIVR'08, Niagara Falls, Canada

Role of MPEG-7 as a way to store metadata
generated for video in a fully standards-based
searchable representation.

Metadata makes digital content searchable (real value is in the metadata!)

MPEG Standards Evolution:

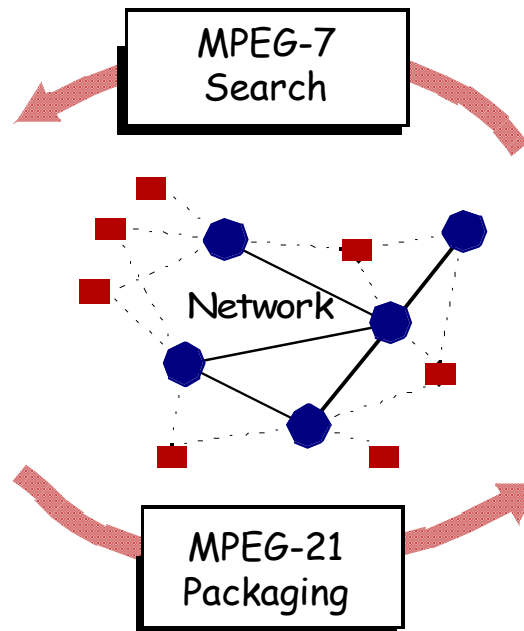
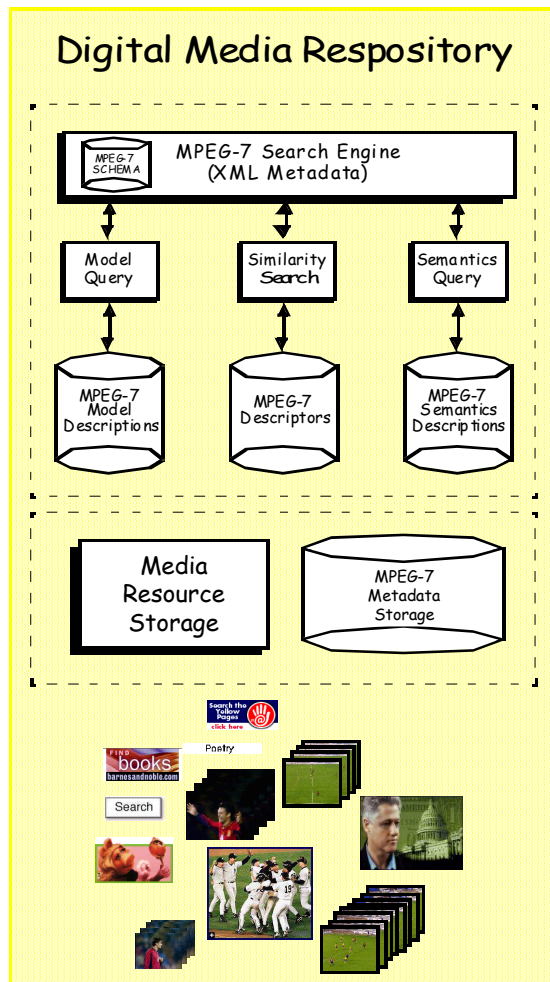
— Coding → “Content Description” → Transactions



MPEG-7/-21 Multimedia Indexing, Searching and Delivery

■ Multimedia Indexing & Searching:

- Semantics-based (people, places, events, objects, scenes, speech)
- Immutable metadata (titles, dates)
- Content-based (color, texture, motion, melody, timbre)

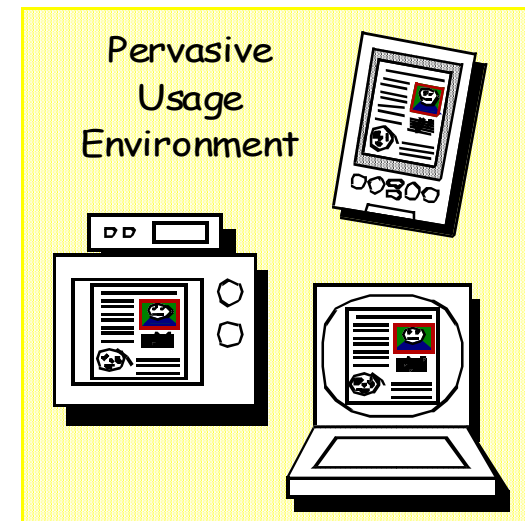
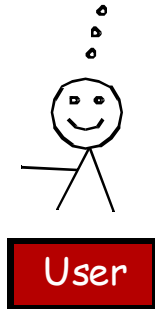


■ Multimedia Access & Delivery:

- Media content personalization
- Adaptation & summarization
- Usage environment (context, devices, user preferences)

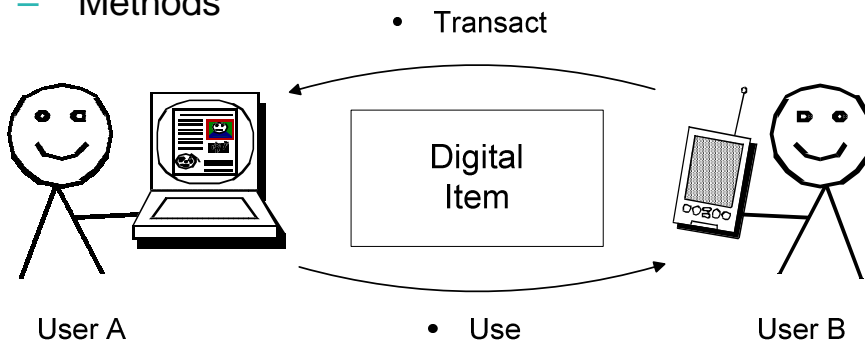


Sounds like ...
Looks like ...

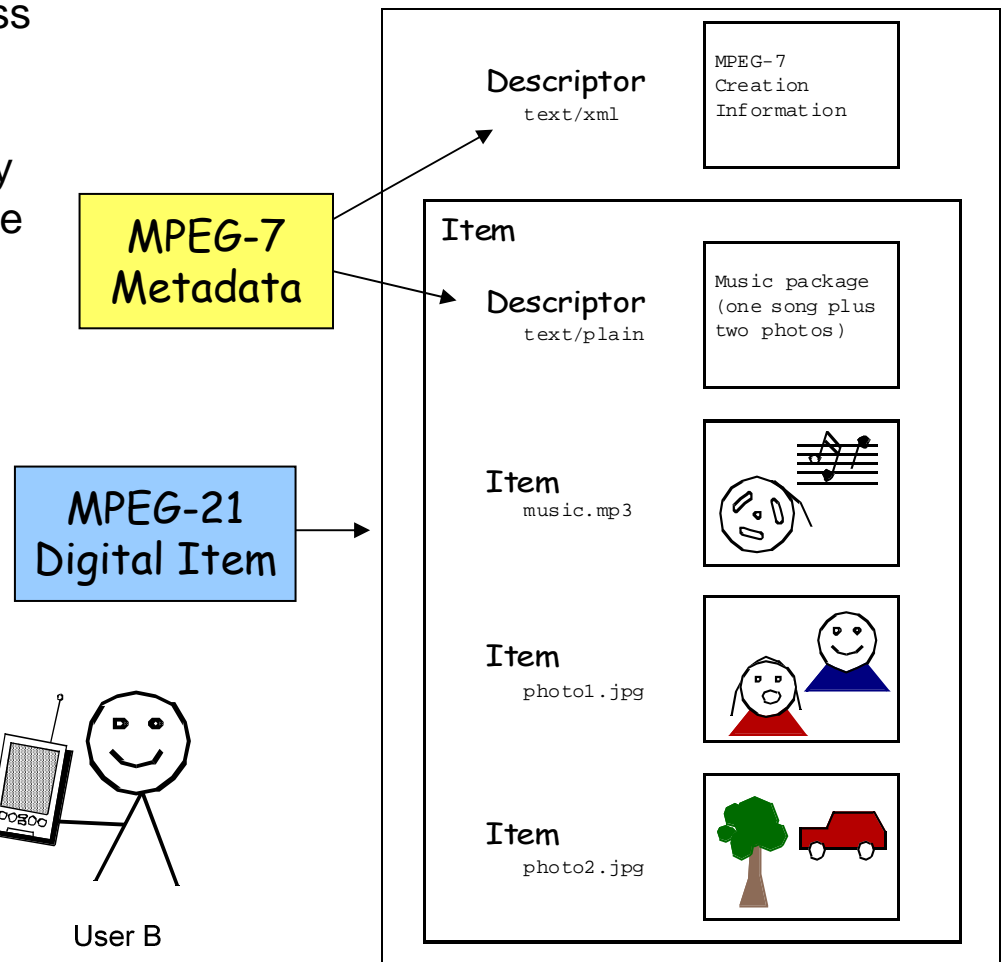


MPEG-21 Multimedia Framework: “Transactions of Digital Items”

- Users and participants in the content value network seamlessly exchange content in form of “digital items” across networks and devices
- Framework supporting all forms of electronic content/intellectual property (video, music, learning objects, on-line reports, etc.)
- Digital Item = bundling of:
 - Media resource
 - Metadata (eg., MPEG-7)
 - Rights expressions
 - Identifiers
 - Methods



- Example: Digital music package



These approaches together go a long way to truly
unleash video search.

References

- **Demos and Tools:**

- IBM Research Marvel “lite”
- <http://www.alphaworks.ibm.com/tech/imars>

- **Links:**

- IBM Research Intelligent Information Management Department:
- <http://www.research.ibm.com/iim>