# VideoCLEF 2008 Pilot

http://ilps.science.uva.nl/Vid2RSS

### Classification of Dual Language **Audio-Visual Content**

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## Goals of VideoCLEF

- Promote research on intelligent access to multimedia content in a multilingual environment
- Encourage exploitation multimodal information streams: speech transcripts, video content, metadata, ...
- Develop and evaluate multilingual video analysis tasks
- Extend the recent Cross-Language Speech Retrieval tracks into new challenges
- Be distinct from TRECVid

## VideoCLEF Vid2RSS Task

- Input: Dual language video, including archival metadata and speech recognition transcripts in Dutch and English
- Output: Series of topic feeds (in RSS format) containing videos; one feed per thematic class



## Data I

- 50 dual language videos (30 hours) from The Netherlands Institute for Sound and Vision (Beeld en Geluid)
- Videos are episodes of Dutch television shows, mostly documentaries
- Dutch is the main (matrix) language; English is an embedded language
- Embedded language is spoken mainly by interviewees

## Data II

- Videos are accompanied by Dutch-language archival metadata records
- Metadata includes series title, episode title, description, date of broadcast and other production information
- Speech recognition transcripts in MPEG-7 format supplied by the University of Twente (both Dutch and English transcripts)
- Shot-level keyframes supplied by Dublin City University

# Subtasks of Vid2RSS

### **Classification Task (Main Task)**

- Assign videos to thematic classes using speech recognition transcripts only (required)
- Use combination of metadata and speech recognition transcripts to perform classification

#### **Translation Task**

Translate output RSS-feeds (e.g., into English)

### **Keyframe Extraction Task**

 Select a keyframe to provide a semantic representation of the entire video to be used to depict the video in the feed.

# **FAQs**

#### Why Dual Language Video?

With appropriate access techniques, information seekers can find spoken content in their own language which is embedded in an archive with an unfamiliar matrix language.

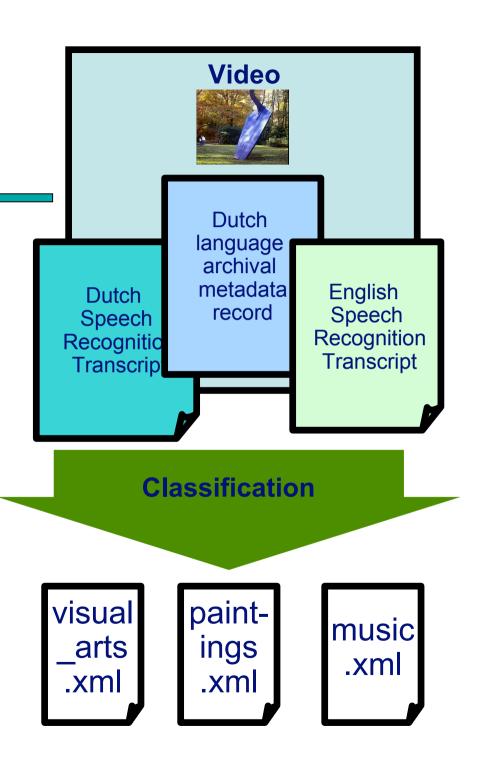
#### Why RSS?

Task results in RSS-format can be directly visualized in a feed reader. They can be immediately assessed by end-users, e.g., archive staff. RSS-format is trivial to generate.

## Classification

#### **Motivation**

- Thematic subject labels encode high-level semantics
- The subject labels have known utility for search
- Ground truth available
- Labels: Archeology,
  Architecture, Chemistry, Dance,
  Film, History, Music, Paintings,
  Scientific research and Visual Arts



# Classification Challenges

- Variation of vocabulary
- Interviewees do not necessarily use topic-specific vocabulary
- Speech recognition errors
- Participants must collect their own training data
- Not a challenge: Feed generation









RunID	micro-averaged	macro-averaged	feature	test doc	site
	f-score	f-score	language	rep	
CUT-C1R1▲	0.15	0.27	en/nl	asr	CUT
CUT-C1R2	0.11	0.14	en/nl	asr	CUT
CUT-C2R1	0.13	0.26	en/nl	asr/md	CUT
CUT-C2R2	0.13	0.17	en/nl	asr/md	CUT
dcu_run1▲	0.41	0.54	nl	asr	DCU
dcu_run2▲	0.25	0.47	en	asr	DCU
dcu_run3▲	0.28	0.58	nl	asr	DCU
dcu_run4	0.28	0.59	en	asr	DCU
dcu_run5	0.29	0.43	nl	md	DCU
MIRACLE-CNL	0.46	0.49	nl	asr	MIRACLE
MIRACLE-CNLEN	0.39	0.27	nl/en	asr	MIRACLE
MIRACLE-CNLMeta▲	0.47	0.47	nl	asr/md	MIRACLE
uams08m	0.18	0.17	nl	md	UAms
uams08asrd	0.10	0.41	nl	asr	UAms
uams08masrd	0.15	0.45	nl	asr/md	UAms
uams08asrde	0.09	0.14	nl/en	asr	UAms
uams08masrde	0.09	0.33	nl/en	asr/md	UAms
SINAI-Class-I	0.51	0.49	nl	asr	SINAI
SINAI-Class-II	0.53	0.51	en	asr	SINAI
SINAI-Class-I-Trans	0.10	0.40	nl	md	SINAI

## Classification: What worked

- Archival metadata and/or Dutch speech recognition transcripts
- Wikipedia, but also general Web, as source of training data
- k-NN/1-NN classifier achieved good precision
- Simplistic retrieval approach: using class labels as queries and video as documents

## Classification: Lessons learned

- Task is not trivial
- Archival metadata and speech recognition transcripts both good feature sources
- Features from speech transcripts of the embedded language (here, English) not helpful
- Performance on certain classes (e.g., Music) was quite acceptable
- Need an evaluation metric that captures human intuitions of performance

## **Translation**

- Carried out by Chemitz University of Technology
- What worked: Google's AJAX language API
- Assessment:
  - Translation evaluated with 3 human assessors rating adequacy and fluency of the translations
  - 2.8 for adequacy (on scale of 1-5)
  - 3.5 for fluency (on scale of 1-5)
- Lesson Learned: Translation of sufficient quantity to make Dutch-language episode descriptions accessible to non-Dutch speaking English speakers

# Keyframe Extraction I

- Carried out by MIRACLE
- Keyframe was selected from set of keyframes provided (one per shot)
- What worked: MIRACLE chose the keyframe whose speech transcript was most representative for the episode

# Keyframe Extraction II

#### Assessment

- 5 human assessors chose keyframe better representative of video episode
- Choice was between manually selected baseline and automatically selected keyframe
- In 44% of the cases, automatic keyframe was chosen by human
- Lesson Learned: Automatic keyframe selection competitive with manual keyframe selection

# Outlook

- Vid2RSS Classification scale-up: more data, more classes
- New tasks under consideration
  - Favorite filtering: Topic independent selection of videos the user's prefer
  - Non-Dutch quote retrieval: Mining the collection for statements useful to non-Dutch speakers
  - Personalized keyframe selection: Choosing representative keyframes most useful to a particular information seeker
  - Finding related resources: Identifying information from non-Dutch sources to support understanding of the video
- Please join us at Friday's breakout session!