Recognizing objects and scenes in news videos

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 Goal: effective retrieval and analysis of large video collections
 Solution required: labeling of objects and scenes
 Problems: manual annotation is not practical recognition on the large scale is still a challenge
 Our Approach: Translation of visual elements to words for

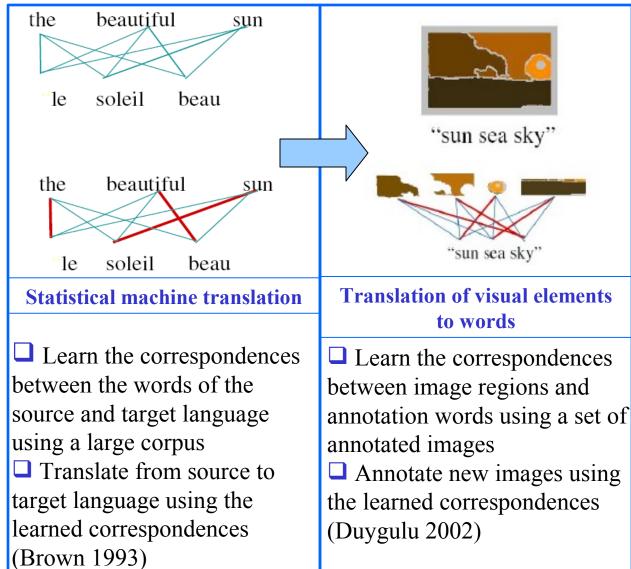
 recognition of objects and scenes on the large scale

- automatic annotation
- better retrieval and analysis

Data Set:

- TRECVID 2004 news videos (~ 350 videos, each ~ 300 keyframes)
- ✤ 114 videos manually annotated by TRECVID participants
- shot & story boundaries, key-frames and speech transcripts are provided by NIST

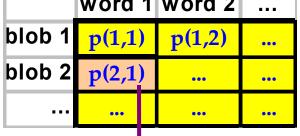
Statistical machine translation methods are adapted



Procedure:

Extract features from video frames
 Vector quantize the features using k-means to transform into discrete elements (blobs)
 Learn the associations between the blobs and words using Giza++ to

obtain a probability table: word 1 word 2 ...



Probability that blob2 matches with word1, P(word1 | blob2)

Use this table to label new images, image regions, etc.

Translating visual elements to words

- similar approach is applied to automatically annotate video frames and to label regions
- however, manual annotations are not always available
- <u>Alternative</u>: use speech transcript text associated with the videos
- **<u>Problem</u>:** speech transcript text is aligned with the shots on the time basis and usually does not correspond to visual information
- **Solution:** learn the correspondences between video frames and speech transcript text inside the news stories



Story 1: weapon inspector iraq president saddam secretary ...

Story 2: air plane transportation ...

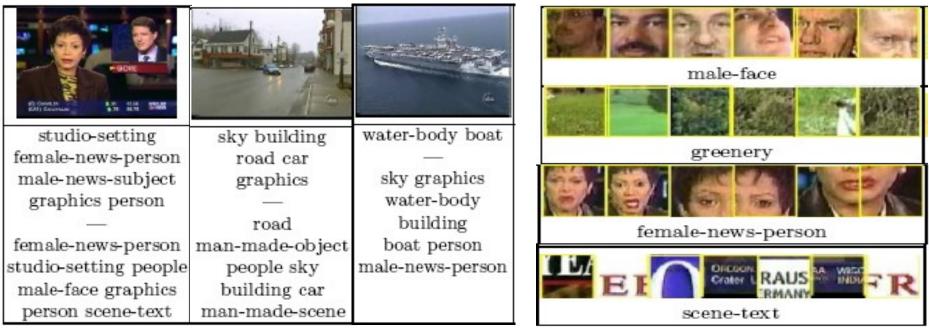
Story 3: game final goal ...

Story 3: weather ...

News stories and associated preprocessed speech transcript text

Labeling results using manual annotations

Data: training set : 10164 images, test set : 7013 images from TRECVID2004 Features: HSV, RGB mean-std, texture (Gabor, Canny) from 5X7 grids; SIFT descriptors, color features around keypoints Keywords: 62 keywords corresponding to nouns



Auto-annotation examples (top : actual, bottom : predicted) **Region labeling examples**

<u>Annotation Performance</u> Average annotation prediction performance per image : 30% For words that are predicted at least once, average recall 18%, average precision 33%

Retrieval results using manual annotations



Queries are performed for all the keywords and first 10 images that are retrieved are examined manually

1st 10 images	MAP (%)
62 words	63
best 30 words	89
best 15 words	99

Ranked query results for: weather-news, cartoon, meeting-room-setting, basketball, female-news-person, food, monitor

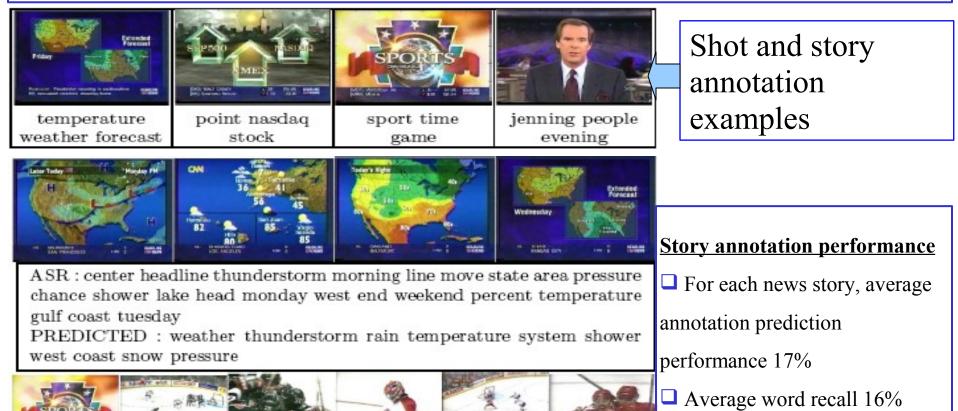
Labeling results using speech transcript text

<u>Data</u>: 31450 frames for training, 31464 frames for test
 <u>Features</u>: global HSV, RGB, edge histograms, SIFT descriptors
 Speech transcripts are in free text form and needs preprocessing. Apply tagging, stemming and stop-word elimination to obtain 251 keywords corresponding to nouns

PREDICTED : game headline sport goal team product business record

ASR : night game sery story

time shot



Average word precision 20%

Retrieval results using speech transcript text



baseball [1,3,4,6,8,9,12]

Ranked query examples [numbers show the ranks]

Future Work

- novel approach to naming many faces by learning the correspondences between the names and faces
- using motion information inherent in video data
 - moving objects are usually more important
 - Use motion information to segment the moving objects
 - learn correspondences between motion information and motion verbs for naming actions and hence supply richer query capability