

# News Image Annotation on a Large Parallel Text-Image Corpus

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LREC - 05/20/2010 - Malta

# Image annotation

Image annotation: textual description of image content



- ← → Noël Forgeard
- ← → Thomas Enders
- ← → Airbus A380
- ← → EADS
- ← → Paris

# Image annotation: major issues



→ Noël Forgeard  
→ Thomas Enders  
→ Airbus A380  
→ EADS  
→ Paris

- How to acquire annotation keywords?
- How to associate keywords with visual content?

# Image annotation: approaches

- Typical approach: low-level approach
  - Fixed annotation vocabulary
  - Visual content = low-level features (color, texture...)
  - Machine learning to find keywords-visual features associations
  - Problem: semantic gap
- Our approach: high-level approach
  - Open annotation vocabulary acquired on a text corpus
  - Detection of high-level concepts in images
  - Association of high-level textual and visual concepts
  - Requires a parallel text-image corpus

# Contributions

- High-level image annotation method
  - Named entity extraction from texts
  - Use of named entities to annotate the images containing the suited visual concepts
- New corpus containing texts and images

# Outline

## 1 Image annotation

- Selection of annotation candidates
- Annotation using visual concepts

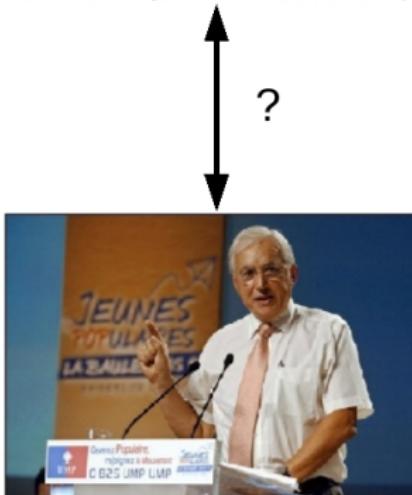
## 2 Text-Image News corpus

## 3 Experiments

## 4 Conclusion

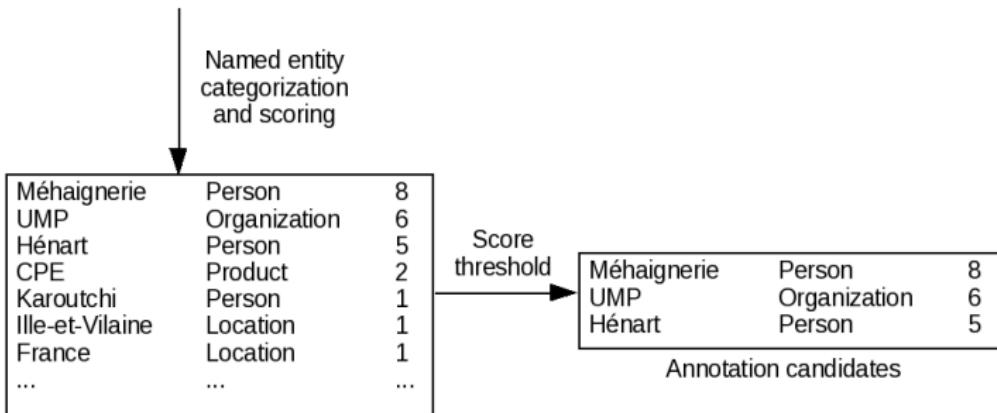
# Objective

PARIS (AFP) - 16/03/2006 19h31 - La grande majorité des élus UMP, déterminés à soutenir un gouvernement avec lequel ils se sentent "sur le même bateau", étaient relativement soulagés jeudi en fin de journée après une manifestation anti-CPE qu'ils jugeaient "moins importante" que prévue. Selon la police, 247.500 personnes, principalement des étudiants et des lycéens, ont manifesté jeudi en France contre le contrat première embauche. "C'est moins que le 7 février", juge, visiblement soulagé, Pierre Méhaignerie, député d'Ille-et-Vilaine et secrétaire général de l'UMP. "Je pensais qu'il y aurait davantage de manifestants", renchérit Laurent Hénart, son collègue de Meurthe-et-Moselle et ancien secrétaire d'Etat à la Formation professionnelle des jeunes. Même réaction de la part de Roger Karoutchi, sénateur des Hauts-de-Seine. "Il y a du monde mais ce n'est pas ce rassemblement massif de jeunes



# Selection of annotation candidates

PARIS (AFP) - 16/03/2006 19h31 - La grande majorité des élus [UMP], déterminés à soutenir un gouvernement avec lequel ils se sentent "sur le même bateau", étaient relativement soulagés jeudi en fin de journée après une manifestation anti-[CPE] qu'ils jugeaient "moins importante" que prévue. Selon la police, 247.500 personnes, principalement des étudiants et des lycéens, ont manifesté jeudi en [France] contre le contrat première embauche. "C'est moins que le 7 février", juge, visiblement soulagé [Pierre Méhaignerie], député d'Ille-et-Vilaine et secrétaire général de l'[UMP]. "Je pensais qu'il y aurait davantage de manifestants", renchérit [Laurent Hénart], son collègue de Meurthe-et-Moselle et ancien secrétaire d'Etat à la Formation professionnelle des jeunes. Même réaction de la part de [Roger Karoutchi], sénateur des Hauts-de-Seine. "Il y a du monde mais ce n'est pas ce rassemblement massif de ...



- Named entity detection and categorization using Nemesis [Fourour, 2002]

## Annotation candidates scores

- Annotation candidates = named entities extracted from the article's text
- Two criterias:
  - Term frequency  $f$ : number of occurrences in the text
  - Document frequency  $df$ : number of documents where the term occurs
- Possible scores:
  - Frequency  $f$
  - Frequency and document frequency  $f - df$
  - Frequency and inverse document frequency  $f - idf$
- Other criterias (see paper)

# Annotation using visual concepts

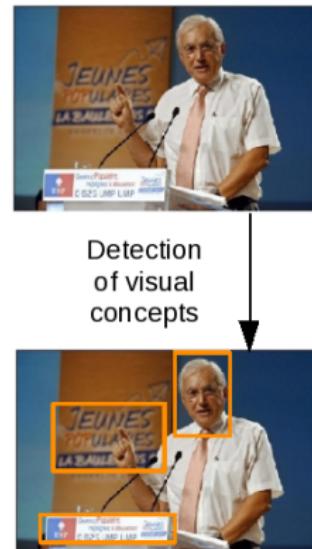


## Annotation candidates

|             |              |   |
|-------------|--------------|---|
| Méhaignerie | Person       | 8 |
| UMP         | Organization | 6 |
| Hénart      | Person       | 5 |

# Annotation using visual concepts

- Face detection: OpenCV detector based on Haar features and boosted decision trees [Lienhart, 2002]
- Logo detection: detector based on visual words and Bayesian learning [Tirilly, 2010]



# Annotation using visual concepts

Annotation candidates

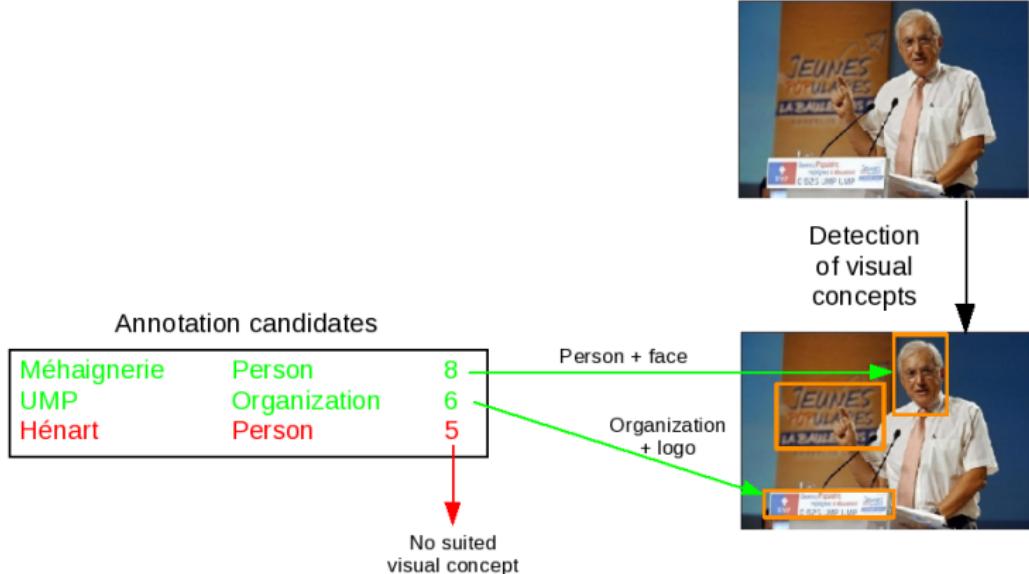
|             |              |   |
|-------------|--------------|---|
| Méhaignerie | Person       | 8 |
| UMP         | Organization | 6 |
| Hénart      | Person       | 5 |



Detection  
of visual  
concepts



# Annotation using visual concepts



# Visual concepts suited to named entities

- Faces ↔ Names of people
- Logos ↔ Names of brands, products, events, organizations, firms, institutions, artistic groups

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# News corpus: document example



# News corpus: properties

- A document contains:
  - an article (text)
  - one or several images
  - one caption per image
  - one context description per image (facultative)
- Corpus properties:
  - Large size (> 27,000 documents, > 42,000 captioned images)
  - Real-world application (news articles)
  - Texts illustrated by images (not only image descriptions)

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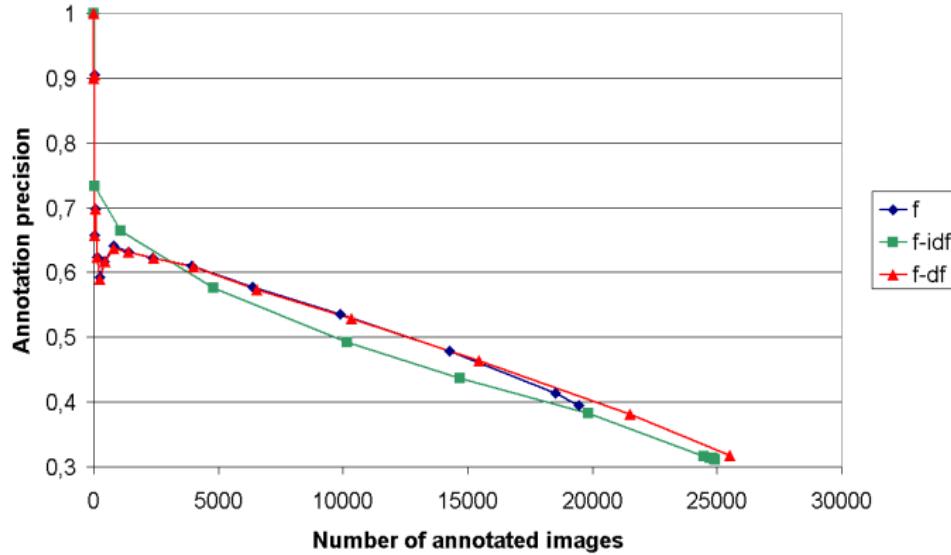
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# Groundtruth and evaluation metrics

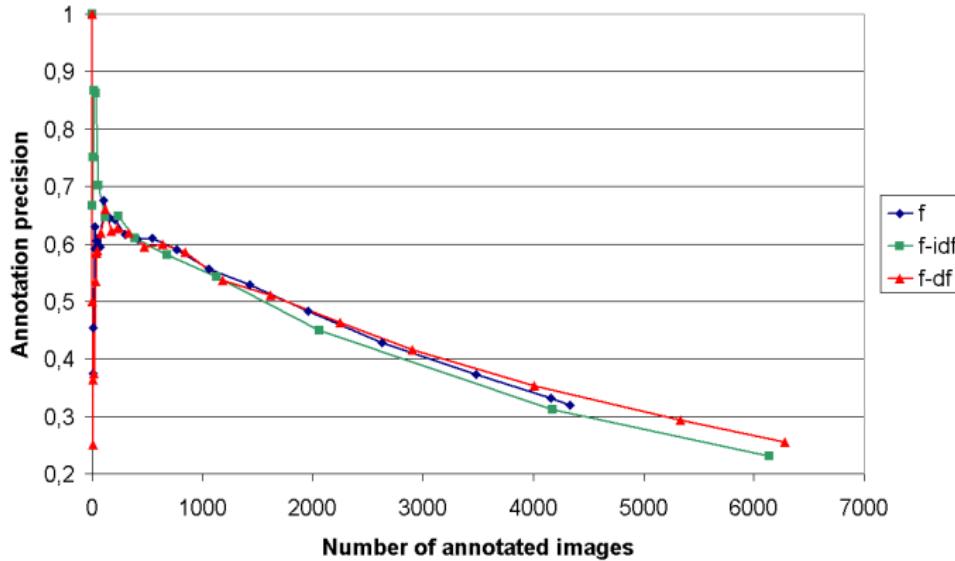
- Groundtruth = captions of images
  - a correct annotation appears in the image's caption
- Evaluation measure = annotation precision (rate of correct annotations) and number of annotated images
  - one threshold to select annotation candidates → one couple (*number of annotated images, annotation precision*)
  - various thresholds → recall-precision-like curves

# Face annotation performance



- $f - idf$  not suited
  - images tend to content common faces

# Logo annotation performance

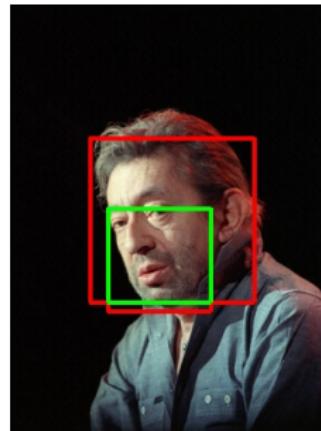


- $f - idf$  not suited
  - images tend to content common logos

# Annotation examples



**Libération** 8  
CE 2  
SCPL 2  
Rotschild 2  
Société Civile des  
Personnels de Libération 1  
Le Monde 1  
Comité d'Entreprise 1



**Gainsbourg** 17  
Nelson 2  
Melody 2  
Birkin 2  
Hardy 2

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# Conclusion

- New bimodal corpus
  - Texts and illustrative images
  - Large-scale and real-world application
- High-level approach to image annotation
  - Open annotation vocabulary (named entities)
  - NLP to acquire annotation vocabulary from texts
  - Use of high level visual concepts
  - Simple technique, but provides good results

## Future work

- Improved named entity selection
  - Length normalization
  - Relative threshold
- Use of other NLP techniques
  - Anaphora resolution
  - Acronym identification
  - Named entity annotation
- Extending our results to other corpora
  - Other news sources
  - Blogs, Wikipedia...

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