

### A contrastive Approach to Multi-word Term Extraction from Domain-specific Corpora

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## Outline

- Introduction and aims
- Multiword extraction process
  - Multi-word candidates extraction
  - Contrastive Re-ranking of extracted terms
- Case studies
- Evaluation
- Conclusions

### The aim

### • Extraction of domain specific terminology \_Focus on multiwords.

### \_Filtering terms from noise:

- Open-domain terms, e.g.
  - \_anno successivo "following year"
- For multi-domain terminology: Singling out terms which belong to different domains.
  - This is the case in the legal domain
  - e.g. environmental terms from legal terms *Rifiuto pericoloso "dangerous waste"*  singled out from *Diritto nazionale "national law"*

## Approaches to Terminology extraction

- The state of the art of TE proposes a wide variety of approaches:
  - Linguistic
  - Statistical
  - Mixed (Linguistic and statistical)
  - **Contrastive**
- Statistical approaches based on e.g. term frequency/inverse document frequency, log likelihood, mutual information, up to more sophisticated approaches such as C-NC Value
- Contrastive approaches: usually applied on single terms extraction
  - They face multiwords extraction expanding the single terms heads
    - e.g. [Basili et. al 2001]: contrastive selection via Heads (CsvH)
    - Single candidate terms selection using a **contrastive function**: distribution in the target and contrastive corpora;

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• Multi-words ranking.

## Our approach: main features

- Multiwords based: we consider multiwords as unique elements, independent from single terms
- **Combines** different approaches: linguistic + statistical + contrastive
- Multi-layered approach
  - We split the multiword extraction process in two steps:
    - Extraction of well-formed multi-word candidates' shortlist
    - Multi-word re-ranking.
  - Benefits of two-step approach:
    - Overcomes the multi-word term sparseness problem
    - Multi-word contrastive ranking: independent from single terms ranking.

## **General workflow**



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### Step 1: Multi-word terms (MWT) candidates extraction

### MWT Candidate extraction process:

- Linguistic filters
  - Based on automatic POS tagged/lemmatized text
  - We identify sequences of allowed POS patterns in order to cover most of the Italian morphosyntactic multi–words structures:
  - Noun+(Prep+(Noun|ADJ)+ |Noun|ADJ)+
    - Diritto nazionale "national law"
    - Presidente della Repubblica "President of the Republic"
  - Filtering of domain specific multi-word preposition, automatically extracted with a first run of the same process using the patterns
  - Noun-Prep-Noun
    - ai sensi di "by law"
- Statistical filters: C-NC Value (Frantzi & Ananiadou 1999).

## Step 2: MWT contrastive ranking

- Candidates multi-word terms are re-ranked using a contrastive method against a reference corpus.
  - i. Single domain contrast against open domain corpus [for filtering noisy general terms]
  - ii. Double domain contrast corpus sharing only one of the domains [for singling out different term types in multi-domain corpora]
  - In case i) TFITF contrastive function
    - Basili et. al 2001 approach: contrastive selection via Heads (CsvH).
    - Our approach: Basili et. al 2001 function directly applied to multiwords.
  - In case ii) CSmw contrastive function
    - based on arctan.
    - Particularly suitable for dealing with low frequency events

## Step 2: MWT contrastive ranking - *TF-ITF contrastive function*

#### **TFITF: Term frequency Inverse Term frequency**

- Variant of Basili et al. 2001
- applied to multi-word terms without passing through single head terms
- A list L of candidate multi-words is extracted with C-NC Value
- L toplist is ranked on TF-ITF score

 $TFITF = \log(f_i(t)) * IWF(t)$ 

- Where  $f_i(t)$  is the frequency of the candidate term (multi-word) *t*, and *IWF* is the inverse word frequency.

 $IWF(t) = \log(N/F(t))$ 

- N: sum of all F(t) for each t in L
- F(t): t frequency in the contrastive corpus

# Step 2: MWT contrastive ranking - *Csmw contrastive function*

#### **Csmw : Contrastive Selection of multi-word terms**

- Specifically designed for dealing with low-frequency events
- Arctan function's mathematical features suites the low frequency events extraction

The statistical weight is calculated directly on multi-word terms

 $CSmw(t) = \arctan(\log(f_i(t)) * K)$ 

Where:  $K(t) = 1/F_c(t)/N_c$ 

C is the set of contrastive domains,  $F_{c}(t)$  is the frequency of t in all contrastive domains of C normalized on N<sub>c</sub> that is the sum of all frequencies in C for each t in L.

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## Case Studies

- Art History case study:
  - Aim: domain specific term extraction
  - Corpus of Art history websites, 326,066 tokens.
    - Manually collected by a domain expert
  - Open domain contrastive Corpus: PAROLE.
    - Italian texts of different types, 3 millions tokens.
- Legislative-environmental case study
  - Aim: "double" domain terminology extraction and classification.
  - Collection of Italian European Legal Texts concerning the environmental domain, 394,088 tokens
  - Contrastive corpora used:
    - PAROLE.corpus.(open-domain)
    - Collection of European Legal Texts concerning the consumer protection domain, 72,210 tokens (Domain specific)

### Art History case study

- Extraction of MWT candidates [C-NC-Value]
  - Selection of a top list of C-NC-Value ranked candidates (threshold empirically set at 600 terms).
- Contrast : against the open domain corpus PAROLE.
  - Final list *L* of 300 domain specific terms.



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## Legislative Case study

- Extraction of MWT candidates [C-NC-Value]
  - Selection of a top list of C-NC-Value ranked candidates (threshold empirically set at 600 terms).
- 1th contrast : against the open domain corpus PAROLE.
  - List *L* of 300 legal and environmental terms.
- 2th contrast: against Legal Corpus on consumer protection
  - Final list *L* new ranking:
    - Top list: <u>environmental terms</u> [rifiuto pericoloso dangerous waste]



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## Evaluation methodology

- Automatic evaluation using gold standard resources
  - Term list provided by domain experts.
  - EartH, Environmental Applications Reference Thesaurus
- Manual evaluation, of unmatched terms, carried out by a domain expert
  - Gold standard resources do not have proper coverage of complex terms.
  - Art domain Art History Department, University of Pisa.
  - Environmental Institute of Atmospheric pollution (CNR).
- Legal Scuola Superiore Sant Anna, Pisa (Ossevatorio sul danno alla persona)
  Evaluation has been carried on wrt the results obtained with:
  - NC-Value
  - Csmw
  - CsvH
  - TF-ITF

### Evaluation – Art history domain

- List of 300 extracted artistic terms
- Extracted MWT distributed into 10 groups of 30 terms each.
- Out of the first 300 terms, CsvH method retrieved the largest amount of Artistic terms.
- TFITF and Csmw have

more domain-specific terms in the top list .

Group	NC-Value	CS-vH	TFITF	Csmw
0-30	24	28	25	25
30-60	20	21	25	24
60-90	20	23	26	25
90-120	18	20	21	24
120-150	20	24	22	26
Tot	102	116	119	124

### Evaluation – Legal domain

- List of 300 extracted artistic terms
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- Top list: mainly environmental terms
- Bottom list: mainly legal terms



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## Conclusions and future developments

Novel approach to MWT extraction combining the C–NC value method with a contrastive ranking technique, aimed at:

- Reducing noise deriving from common words
- Discriminating semantically different types of terms within heterogeneous terminologies (as in the legal domain)

### - Current directions of research include:

Improvements to the MWT extraction algorithm

Improvements of the multi-domain terminology extraction task

Application of the proposed approach to identify neologisms from diachronic corpora of newspapers texts.

Thanks for your attention!