

# Extracting MWEs from Italian corpora: A case study for refining the POS-pattern methodology

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## Abstract

An established method for MWE extraction is the combined use of previously identified POS-patterns and association measures. However, the selection of such POS-patterns is rarely debated. Focusing on Italian MWEs containing at least one adjective, we set out to explore how candidate POS-patterns listed in relevant literature and lexicographic sources compare with POS sequences exhibited by statistically significant n-grams including an adjective position extracted from a large corpus of Italian. All literature-derived patterns are found—and new meaningful candidate patterns emerge—among the top-ranking trigrams for three association measures. We conclude that a final solid set to be used for MWE extraction will have to be further refined through a combination of association measures as well as manual inspection.

## 1 Introduction

The CombiNet project<sup>1</sup> has the goal of building an online resource for Word Combinations in Italian, including MWEs of various degrees of fixedness (such as phrasal lexemes, collocations and usual combinations) as well as distributional profiles of Italian lexemes. Within this project, the present paper aims at investigating ways to refine a well-known methodology for MWE-extraction, namely the combined use of previously identified POS-patterns and association measures (Evert and Krenn, 2005). While POS-patterns are widely used to extract MWEs from corpora in order to constrain the array of possible outputs (Krenn and Evert, 2001; Wermter and Hahn, 2006, e.g.), the way in which POS-patterns are created in the first place is much less addressed. This step is however crucial,

<sup>1</sup><https://sites.google.com/site/enwcin/home>

especially considering that the list of patterns is necessarily language-specific. The goal of this paper is to propose a method to optimize – in terms of both recall and precision – the list of POS patterns to be used for the subsequent extraction of potential MWEs. In order to do this, we compare predetermined patterns, which would be normally used as a first-pass sieve for potential MWEs, with patterns exhibited by statistically significant n-grams extracted from data.

## 2 Methodology

In this pilot study, we focus on MWEs containing at least one adjective, and we limit the extraction to trigrams (Section 2.1). We make use of the following sets of data: (a) a list of frequently used Italian adjectives; (b) a list of previously identified POS-patterns containing at least one adjective.<sup>2</sup>

The adjectival lemmas were taken from the *Senso Comune* dictionary,<sup>3</sup> which contains 2,010 fundamental lemmas of the Italian lexicon, 211 of which are adjectives (e.g. *bello* “beautiful”, *brutto* “ugly”, *ricco* “rich”). These adjectives are used to constrain the extraction procedure, and we refer to this set as  $\{SC\}$ .

The list of predetermined POS-patterns for MWEs involving one adjective was obtained by merging the following information: (a) patterns of word combinations included in existing combinatory dictionaries for Italian (Pianno et al., 2013), see Table 1a; (b) additional combinatory types mentioned in the relevant theoretical literature (Voghera, 2004; Masini, 2012), summarised in Table 1b; and (c) a few more patterns based on our own intuition, i.e. identified by elaborating on the previous two lists (Table 2). This joint collection contains a total of 19 patterns, most of which are bigrams (11), and fewer are trigrams (8). Note

<sup>2</sup>For information on POS tags see Appendix.

<sup>3</sup><http://www.sensocomune.it/>

that trigrams (put together in Table 2) come for the most part from our intuition, indicating that these patterns are rather neglected in the literature and in combinatory dictionaries of Italian, which tend to focus on bigrams. For this reason, and because longer sequences are intrinsically more idiosyncratic, we concentrate on trigrams for this pilot experiment, although in the discussion we take into account bigrams, too (Section 3).

Table 1: Italian POS-patterns with ADJ(s)

POS-pattern	Example	Translation
(a) from lexicographic sources		
ADJ ADJ	stanco morto	dead tired
ADJ CON ADJ	vivo e vegeto	live and kicking
ADJ NOUN	prima classe	first class
ADJ PRE	pronto a	ready to
ADV ADJ	molto malato	very ill
NOUN ADJ	casa editrice	publishing house
VER ADJ	uscire pazzo	to go crazy
(b) from relevant literature		
ADJ PRO	qual esso	which/who
ADJ <sub>i</sub> ADJ <sub>i</sub>	papale papale	bluntly
ARTPRE ADJ	alla francese	French-style
PRE ADJ	a caldo	on the spot
PRE ADJ NOUN	di bassa lega	vulgar/coarse
PRE NOUN ADJ	a senso unico	one-way
PRO ADJ	tal altro	some other

## 2.1 Extracting the trigrams

From the corpus La Repubblica (Baroni et al., 2004), which consists of 300M words of newswire contemporary Italian, we extracted all trigrams featuring at least one adjective, deriving this information from the pre-existing POS tags in the corpus. All trigrams were extracted as sequences of lemmas. We created three separate lists according to the adjective’s position in the trigram (first, second, or third). All instances containing any punctuation item were discarded.

For each of the three sets, we kept only trigrams occurring more than five times in the whole corpus. As a further step, we selected those instances featuring one of the 211 adjectives in  $\{SC\}$ , yielding a total of 89,217 different trigrams featuring an adjective as first member (191 adjectives from  $\{SC\}$  occur in this position), 100,861 as second (192 adjectives), and 114,672 as third (193).

## 2.2 Ranking the trigrams

We used the Text-NSP package (Banerjee and Pedersen, 2003) to rank the trigrams in each of the

three sets according to three association measures (AMs), namely the Poisson-Stirling measure (PS), the log-likelihood ratio (LL) and pointwise mutual information (PMI). However, on the basis of preliminary inspection and observations in the literature on ranking Italian MWEs extracted from corpora (Nissim and Zaninello, 2013), we discarded PMI as not too accurate for this task. We also considered raw frequencies, as they have proved good indicators for collocations, on a par with AMs (Krenn and Evert, 2001; Bannard, 2007).

The idea is to check which POS sequences are exhibited by the highest instances in the rank, under the rationale that such patterns might be good representations of Italian MWEs containing adjectives, and can be used for further extraction and characterisation of the phenomenon (in dictionaries and resources). Thus, we selected the top 10% instances in each rank, extracted their POS patterns, and ranked such patterns according to the number of times they appeared. Tables 3–5 report the ten most frequent patterns according to each measure, when an adjective is found in first, second, and third position, respectively.

## 3 Analysis and discussion

By comparing the ranked patterns in Tables 3–5 with the predetermined POS-patterns for trigrams in Table 2, we draw the following observations.

We first consider patterns that are ranked high for *all* measures. Some find a correspondence to those in Table 2, implying that these are likely to be solid, characteristic POS sequences to be used in extraction (ADJ CONJ ADJ (for ADJ in first position), ADJ PRE VER, PRE ADJ NOUN, and VER PRE ADJ). Other found patterns, instead, are *not* among the pre-identified ones, but are definitely typical sequences, as the analysis of some of the extracted trigrams shows. Among these: ADJ PRE NOUN (*ospite d’onore* “special guest”), VER ART ADJ (*essere il solo* “to be the only one”), NOUN PRE ADJ (*agente in borghese* “plain-clothes policeman”), ARTPRE ADJ NOUN (*all’ultimo momento* “at the last moment”). Envisaging an extraction procedure based on POS sequences, such structures should be included to improve recall.

Conversely, the PRE ART ADJ pattern exhibits an incomplete sequence, and is therefore unsuitable for MWE extraction. Since the inclusion of such patterns would possibly affect precision, they need to be filtered out on the grounds of grammatical

Table 2: Trigram POS-patterns containing ADJ(s)

POS-pattern	Example	Translation
from literature and resources		
ADJ CON ADJ	pura e semplice	pure and simple
PRE ADJ NOUN	a breve termine	short-run
PRE NOUN ADJ	in tempo reale	(in) real-time
from our own intuition		
ADJ PRE VER	duro a morire	die-hard
NOUN ADJ ADJ	prodotto interno lordo	gross national product
NOUN NOUN ADJ	dipartimento affari sociali	social affairs division
PRE ADJ VER	per quieto vivere	for the sake of quiet and peace
VER PRE ADJ	dare per scontato	to take for granted

Table 3: Top 10 POS patterns featuring an adjective as word1, extracted from the top 10% trigrams ranked according to LL, PS, and raw frequency.

LL	PS	raw frequency
ADJ PRE VER	ADJ NOUN PRE	ADJ NOUN PRE
ADJ PRE ART	ADJ NOUN ARTPRE	ADJ NOUN ARTPRE
ADJ NOUN PRE	ADJ NOUN ADJ	ADJ ARTPRE NOUN
ADJ PRE NOUN	ADJ ARTPRE NOUN	ADJ PRE ART
ADJ NOUN ARTPRE	ADJ PRE VER	ADJ PRE VER
ADJ ARTPRE NOUN	ADJ PRE NOUN	ADJ NOUN ADJ
ADJ PRE DET	ADJ CON ADJ	ADJ PRE NOUN
ADJ CON ADJ	ADJ NPR NPR	ADJ CON ADJ
ADJ CHE CLI	ADJ NOUN CON	ADJ NOUN CON
ADJ DET NOUN	ADJ PRE ART	ADJ CON ART

constraints, or, ultimately, manual inspection.

Additionally, there are patterns that *contain* or are *portions of* more relevant patterns for MWE-hood. Some capture what are in fact bigrams (Table 6), while others are portions of 4-grams or possibly larger sequences, namely NOUN ADJ PRE (NOUN), (NOUN) ADJ ARTPRE NOUN, and NOUN ARTPRE ADJ (NOUN), where the “missing” POS is given in brackets. Examples are: *concorso esterno in (omicidio)* “external participation in (murder)”, *(banca) nazionale del lavoro* “National (Bank) of Labour”, and *paese del terzo (mondo)* “third world (country)”, respectively. Running a full-scale extraction procedure that accounts for all n-grams should naturally take care of this.

Some of patterns from Table 2 are ranked high only by *some measures*: PRE NOUN ADJ only according to PS and raw frequency (Table 5), and NOUN ADJ ADJ both for second and third position, but only by PS. Overall, with respect to their ability to extract previously identified POS-patterns, AMs perform similarly when the adjective

is the first member (Table 3), whereas PS seems to be more indicative when the adjective is second and third (Tables 4-5), together with raw frequency, while LL seems to be generally performing the worst. This point calls for a combination of AMs (Pecina, 2008), but will require further work.

As for predetermined patterns that are *not* found among the top ones, we observe that NOUN NOUN ADJ is basically an adjective modifying a noun-noun compound, and should be best treated as a “complex bigram”. Similarly, the PRE ADJ VER pattern can be seen as an extension of the ADJ VER bigram, which is usually not considered (Table 1). Investigating the combination of bigrams, trigrams and n-grams with  $n > 3$  is left for future work.

## 4 Conclusion

In summary, basically all of the literature/intuition-based patterns are retrieved from highly ranked plain trigrams. However, top-ranking trigrams also exhibit other POS sequences which should be included in a set of patterns used for MWE extrac-

Table 4: Top 10 POS patterns featuring an adjective as word2, extracted from the top 10% trigrams ranked according to LL, PS, and raw frequency.

LL	PS	raw frequency
ART ADJ NOUN	ART ADJ NOUN	ART ADJ NOUN
NOUN ADJ PRE	ARTPRE ADJ NOUN	ARTPRE ADJ NOUN
PRE ADJ NOUN	PRE ADJ NOUN	PRE ADJ NOUN
ARTPRE ADJ NOUN	NOUN ADJ ARTPRE	NOUN ADJ PRE
DET ADJ NOUN	NOUN ADJ PRE	NOUN ADJ ARTPRE
ART ADJ NPR	NOUN ADJ CON	NOUN ADJ CON
ART ADJ CON	DET ADJ NOUN	ADV ADJ PRE
ADV ADJ PRE	VER ADJ NOUN	DET ADJ NOUN
DET ADJ VER	NOUN ADJ ADJ	ADV ADJ ARTPRE
VER ADJ PRE	ADV ADJ PRE	ADV ADJ CON

Table 5: Top 10 POS patterns featuring an adjective as word3, extracted from the top 10% trigrams ranked according to LL, PS, and raw frequency.

LL	PS	raw frequency
VER ART ADJ	ART NOUN ADJ	ART NOUN ADJ
PRE ART ADJ	ARTPRE NOUN ADJ	ARTPRE NOUN ADJ
NOUN PRE ADJ	PRE NOUN ADJ	VER ART ADJ
NOUN ARTPRE ADJ	NOUN ARTPRE ADJ	PRE ART ADJ
VER ARTPRE ADJ	VER ART ADJ	PRE NOUN ADJ
VER PRE ADJ	NOUN PRE ADJ	NOUN ARTPRE ADJ
NOUN ART ADJ	PRE ART ADJ	NOUN PRE ADJ
ADV ART ADJ	NOUN ADV ADJ	VER PRE ADJ
ADV ADV ADJ	VER PRE ADJ	NOUN ADV ADJ
ART DET ADJ	NOUN ADJ ADJ	VER ARTPRE ADJ

Table 6: Extracted trigram patterns that subsume a bigram pattern (boldfaced).

Pattern	Example	Translation
<b>ADJ PRE ART</b>	<b>degn</b> <b>di</b> un	<b>worthy of a</b>
<b>ADJ NOUN PRE</b>	<b>utile netto</b> di	<b>net profit of</b>
<b>ADJ NOUN ARTPRE</b>	<b>alto funzionario</b> del	<b>senior official of</b>
<b>ART ADJ NOUN</b>	il <b>pubblico ministero</b>	the <b>public prosecutor</b>
<b>NOUN ADJ PRE</b>	<b>centro storico</b> di	<b>historical centre of</b>
<b>ARTPRE ADJ NOUN</b>	della <b>pubblica amministrazione</b>	of the <b>public administration</b>
<b>DET ADJ NOUN</b>	altro <b>duro colpo</b>	another <b>hard blow</b>
<b>ADV ADJ PRE</b>	sempre <b>pronto a</b>	always <b>ready to</b>

tion to improve recall. At the same time, several patterns extracted with this technique are to be discarded. Some are just irrelevant (e.g. ADJ CHE CLI, *nero che le* “black that them”): in this respect, combining various AMs or setting grammatical constraints could help refine precision, but human intervention also seems unavoidable. Others are not meaningful trigrams as such, but may be meaningful as parts of larger MWEs or because they contain meaningful bigrams. Here, it would be in-

teresting to explore how to combine n-grams with different n-values.

This pilot experiment shows that trigram ranking is useful to extract new patterns that are not considered in the initial set. The latter can be therefore expanded by following the proposed methodology, as a preliminary step towards the actual extraction of candidate MWEs from corpora. Clearly, the validity of the expanded POS-pattern set can only be evaluated after the extraction step is completed.

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## Appendix

The tagset for all patterns extracted from the corpus “La Repubblica” is accessible at <http://sslmit.unibo.it/~baroni/collocazioni/itwac.tagset.txt>. In the context of this experiment we collapsed all fine-grained tags into the corresponding coarse-grained tag (e.g. all verbal tags such as VER:fin or VER:ppast were collapsed into VER). The POS tags used in this paper are to be interpreted as in the Table below.

Table 7: POS tags used in this paper.

abbreviation	part of speech
ADJ	adjective
ADV	adverb
ART	article
ARTPRE	prepositional article
CHE	any function of the word “che” (adjective, conjunction, pronoun)
CLI	clitic
DET	determiner
NOUN	noun
PRE	preposition
VER	verb