



MWE: I CAN'T HELP FALLING IN LOVE WITH YOU

FEDERICO SANGATI
FBK-DH, Trento
sangati@fbk.eu

ANDREAS VAN CRANENBURGH
Huygens ING, Royal Netherlands Academy of
Arts & Sciences; ILLC, Univ. of Amsterdam.
andreas.van.cranenburgh@huygens.knaw.nl

JOHANNA MONTI
Sassari University
jmonti@uniss.it



FONDAZIONE
BRUNO KESSLER

ABSTRACT

We investigate new ways for identifying MWEs from parallel multi-lingual corpora, based on the **non-translatability** property of MWEs: an MWE cannot be translated from one language to another on a word by word basis (Sag et al., 2002; Monti, 2012).

MWE AND NON-TRANSLATABILITY

TARGET CASES

- Fixed expressions** e.g., EN. by and large → IT. *da e largo.
- Idioms** e.g., EN. Call it a day → IT. *Chiamarlo un giorno.
- Proverbs** e.g., EN. There's no such thing as a free lunch → IT. *Non esiste una cosa come un pranzo gratuito.
- Phrasal verbs** e.g., EN. Bring somebody down → IT. *Portare qualcuno giù.

EXCEPTIONS

A number of MWEs can be translated literally to all other languages, such as **proper names** and **universal proverbs**. These are therefore excluded from the scope of the current work.

PHASE 1: KERNEL METHODS

Goal: Identify potential MWEs in parallel pairs of sentences (in one language, in the other, or in both).

Input: large bilingual corpus sentence aligned.

Kernel methodology:

- For every pair of sentences in the corpus, the algorithm will detect a pair of sentences in the source language which **share** a certain expression, for which the correspondent pair of sentences in the target language also **share** an expression.
- Can be computed efficiently via **string kernel** (Lodhi et al., 2002) for aligned text, while **tree kernels** can be employed (Sangati et al., 2010; van Cranenburgh, 2014) if a parallel treebank is available.

Example:

English

Italian

I feel we will have to **call it a day** at this point.

Credo che a questo punto dobbiamo **passare oltre**.

He would like us to adjourn the vote to the next part-session and **call it a day** for now.

Il relatore chiede di rinviare la votazione alla prossima seduta e, per ora, di **passare oltre**.

Outcome cases:

	English	Italian
1.	MWE bring up to date	×
2.	×	MWE ha tirato le cuoia
3.	MWE call it a day	MWE passare oltre
4.	×	×
	aims at adapting	mira ad adattare

PARSEME WORKING GROUPS

WG1: Lexicon-Grammar Interface Development of linguistic resources, MWE dictionaries.

WG3: Statistical, Hybrid and Multilingual Processing of MWEs Hybrid methodology for MWE identification and translation.

PHASE 2: MT FILTERING

Goal: remove candidate pairs without MWEs.

- Phase 1 is prone to find many pairs of candidate expressions which do not include MWEs (e.g., last row of outcome cases).

Methods:

- Traditional "word by word" translation system (detect which candidate pairs are literal translations).
- 1:1 alignment pairs between source and target languages obtained via GIZA++ (Och and Ney, 2003).

PHASE 3: CROWDSOURCING

Goal: validate the final list of candidate pairs using crowdsourcing methods:

- Amazon Mechanical Turk
- CrowdFlower
- Educ. tools for second language learners
- CAT systems for human translators

RELATED WORK

Some recent approaches rely on the exploitation of the translational correspondences of MWEs.

De Medeiros Caseli et al. (2010) identification of MWEs in a multilingual context, exploiting a word alignment process. Also associates some multiword expressions with semantics.

Tsvetkov and Wintner (2014) exploit non-compositional translation of MWEs and developed a new alignment-based algorithm for MWE extraction focused on misalignments, augmented by validating statistics computed from a monolingual corpus.

Segura and Prince (2014) propose an alignment process between pairs of sentences, strongly based on syntax. It relies on a rule-based system combining partial alignments from a database through a non-iterative graph-theory based process.

Arcan et al. (2014) address the problems of automatic identification of bilingual terminology using Wikipedia as a lexical resource, and its integration into an SMT system using the XML mark-up and the Fill-Up model methods.

CASE STUDY: CAN'T HELP

Corpus: TED Talks EN-IT (Cettolo et al., 2012)

- Number of sentences: 187,809
- Tokenized and aligned with GIZA++ (many thanks to Mihael Arcan)

Target MWE: EN: **can't help** → IT: **fare a meno di**

Corpus Analysis:

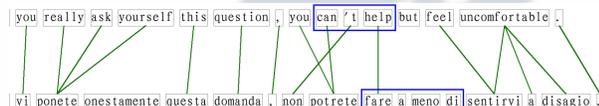
- Intersection** (4) [1760, 41845, 87214, 107792]
- Only in EN** (22) [9303, 9316, 13677, 13687, 15336, 22592, ...]
- Only in IT** (7) [41031, 41213, 46509, 101575, 117009, 161383, 165466]

GIZA++ alignments:

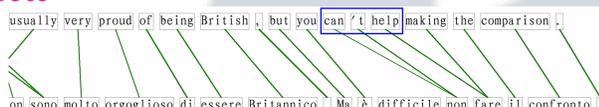
1760



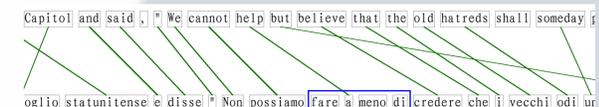
107792



9303



41031



MT Systems:

EN source: "I **can't help** falling in love with you."

Google (2014.09.01) Correct

* Non posso fare a innamorarsi di te.	Non posso fare a meno di innamorarmi di te.
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