

A DISTRIBUTIONAL SEMANTICS APPROACH TO COMPETITION IN ITALIAN DEADJECTIVAL VERB FORMATION

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References

1. Introduction

The study provides some insights into the **competition between derivational processes in deadjectival verb formation in Italian** by means of a corpus based analysis within the frame of Distributional Semantics (DS). Through DS we calculate the general vectors of each process and the similarity values between pairs of co-radical verbs. Our aim is to assess which derivational processes more often yield semantically similar verbs, and if **semantic similarity between verbs assumed on the basis of dictionary definitions is corroborated by the distribution of these verbs in corpora**.

3. Deadjectival verb formation

The derivational processes of **suffixation**, **conversion** (or zero derivation), and **parasynthesis**

- **potentially compete** for the expression of the core meanings to which the semantics of deadjectival verbs can be tracked back to. Specifically, the encoding of
 - **causative/resultative** ('make/become (more) Adj') meanings, i.e. of change of state, is achieved by means of all three processes:
e.g. suff. *inglese* → *inglesizzare* *puro* → *purificare* *amaro* → *amareggiare*
paras. *bello* → *abbellire* *giallo* → *ingiallire* *doppio* → *sdoppiare*
conv. *sano* → *sangare* *scuri* → *scurire*
 - **stative/similative** ('be / act Adj') meanings, i.e. no change of state, is achieved by *-eggiare* suffixation and conversion
e.g. suff. *folle* → *folleggiare* conv. *paziente* → *pazientare*
- have, over time, often been employed to form multiple **verbs from a same adjectival base**: in some cases to express different meanings, in others to encode (fully or near-to) equivalent semantics, giving rise to **lexeme competition**:
e.g. *curvo* → a. suff. *curveggiare* 'to wind'
b. conv. *curvare* 'to bend'
c. paras. *incurvare* 'to bend'

2. Distributional Semantics

The core assumption of Distributional Semantics is that **semantically similar words show a similar distribution in linguistic contexts** (Harris, 1954).

This hypothesis is operationalized by calculating the semantic similarity between two words in terms of **similarity between their vectors**, which are a **statistical representation of the contexts** in which the words appear (Lenci, 2018).

To perform our analyses we employ Word2Vec (w2v) (Mikolov et al., 2013), a widely used neural network model trained on itWaC, a 2-billion-word Italian corpus constructed from the Web (Baroni et al., 2009), and the Alacarte embedding algorithm (Khodak et al., 2018).

4. Data

The dataset includes **784 verbs** (32.7% suffixed, 38.6% parasynthetics, 28.7% converted)

- collected from the *Grande Dizionario della Lingua Italiana*, the reference historical dictionary, and the *Dizionario Italiano Sabatini Coletti*, a dictionary of contemporary usage
- occurring with a frequency > 10 in itWaC, our source for frequencies and vectors
- classified based on derivational process and pattern (suffixes: *-izzare*, *-eggiare*, *-ificare*; parasynthetic patterns: *ad-*, *in-*, *s-*, *de-*, *di-*, *ri-*, *ra-*, *rin-* Adj *-a/ire*; inflectional endings for converted verbs: *-are*, *ire*).

process	n_bases	n_types	n_tokens
conversion	215	225	4,420,657
parasynthesis	217	303	2,043,692
-izzare suffixation	195	196	1,405,112
-eggiare suffixation	35	35	27,099
-ificare suffixation	25	25	476,753

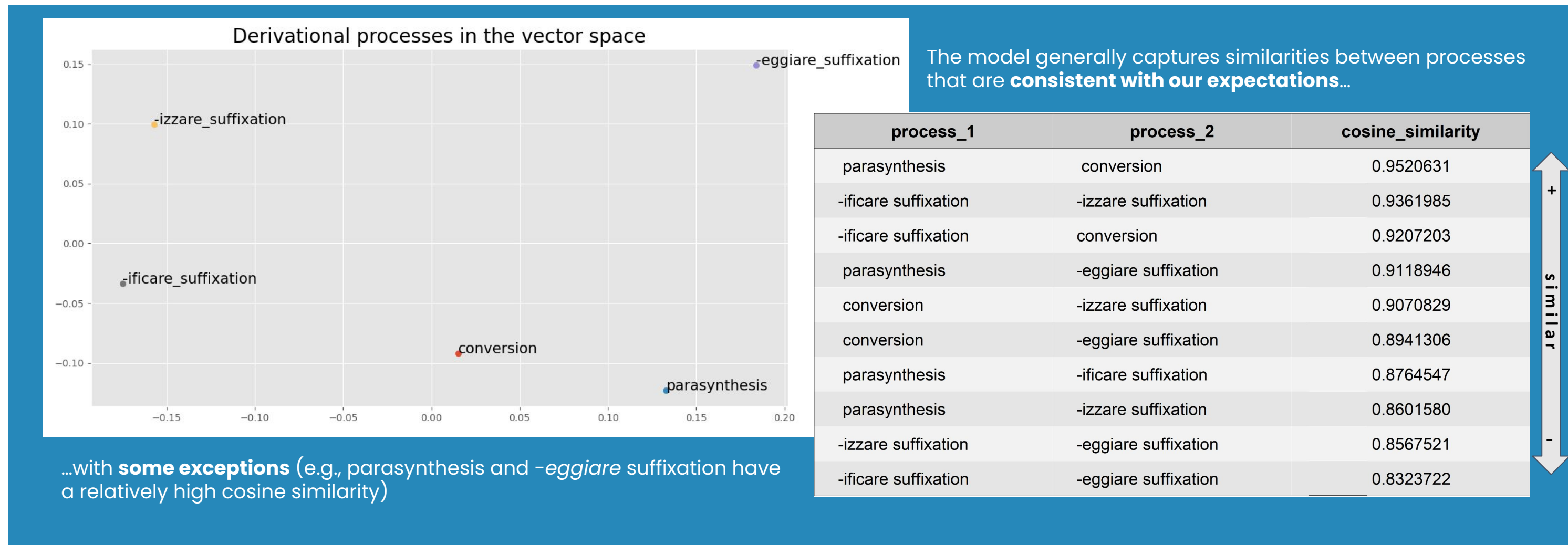
Research questions...

...and how we tackled them

- Which derivational processes appear to be more semantically similar (i.e., form verbs with more similar semantics)?
- How is semantic similarity between processes reflected into semantic similarity between co-radical verbs?
- Under which conditions do synonymous co-radical verbs formed by different processes coexist?

- 1) We generate an average vector for each process and compare them with Cosine Similarity to find the most distributionally similar ones (Guzmán Naranjo & Bonami 2023).
- 2) We generate similarity values between pairs of co-radical verbs. Then we calculate the average of similarities for each pair of processes.
- 3) We analyze the most similar neighbours of co-radical verbs pairs to detect difference in the meanings captured by the DS approach in a finer-grained fashion.

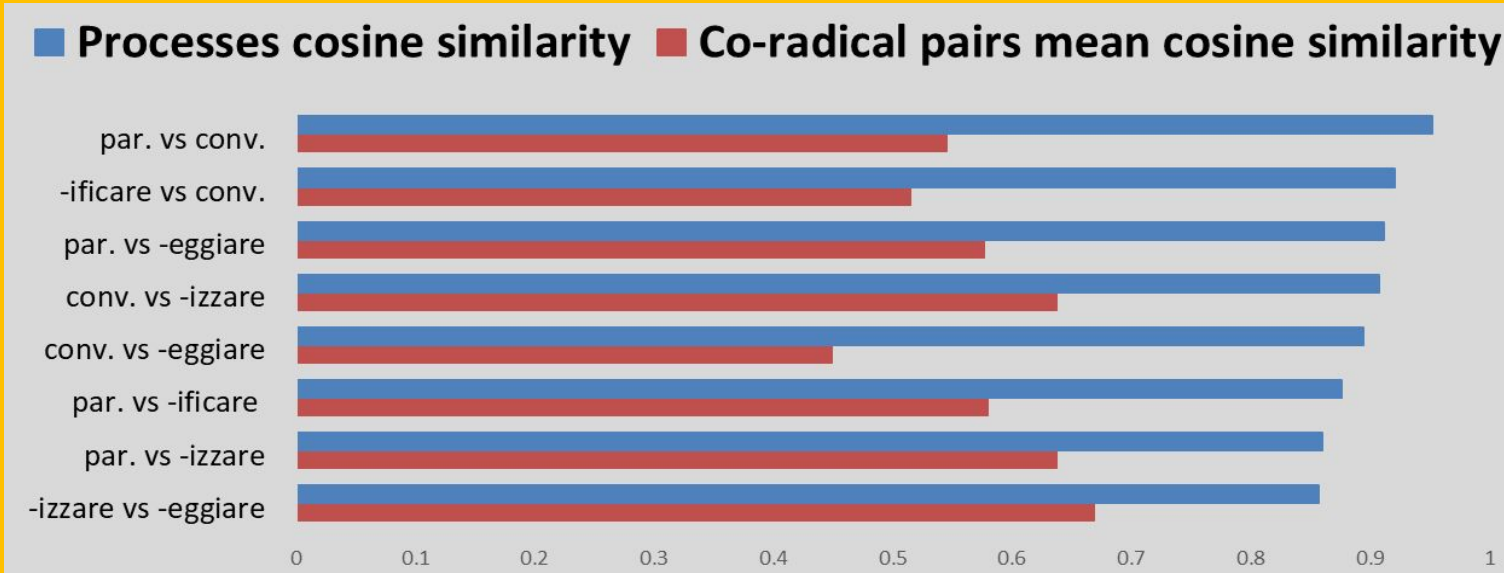
a) Process similarity



b) Co-radicals mean similarity

process_1	process_2	mean_cos_similarity	n_pairs
-izzare suffixation	-eggiare suffixation	0.6688296	9
conversion	-izzare suffixation	0.6376260	10
parasynthesis	-izzare suffixation	0.6369260	8
parasynthesis	-ificare suffixation	0.5790315	23
parasynthesis	-eggiare suffixation	0.5766763	30
conversion	parasynthesis	0.5451411	131
conversion	-ificare suffixation	0.5146794	12
conversion	-eggiare suffixation	0.4493575	11

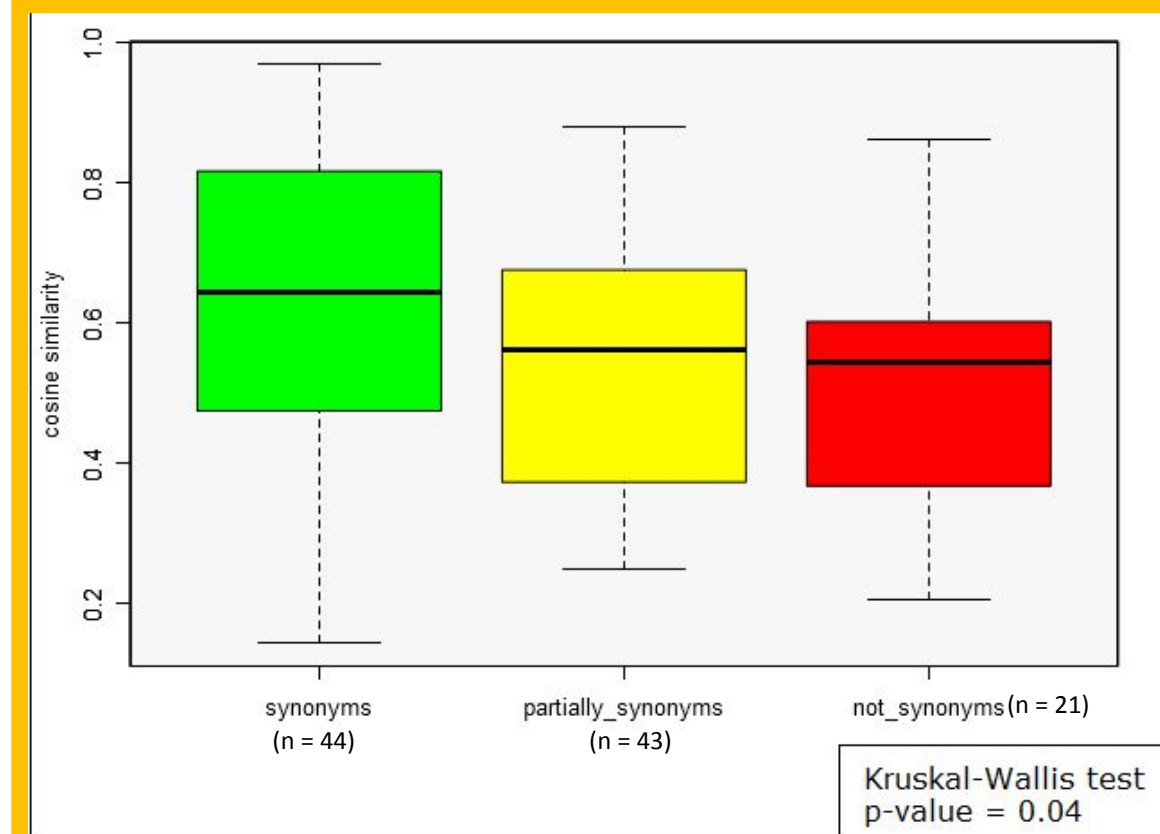
Co-radical **pairs formed by means of the most similar processes** seem to be on average among the **least similar ones** (e.g., see conversion and parasynthesis).



c) Co-radical similarity: The case of converted and parasynthetics

Through a comparison with dictionary definitions (GRADIT), we **assess the similarity scores** assigned to the "conversion vs parasynthesis" pairs of verbs, in order to test the reliability of data shown in b). We assigned three values:

- synonyms: the meanings fully match
- partially syn.: some of the meanings are shared by both verbs, others are specific to each of them
- not syn.: none of the meanings are shared



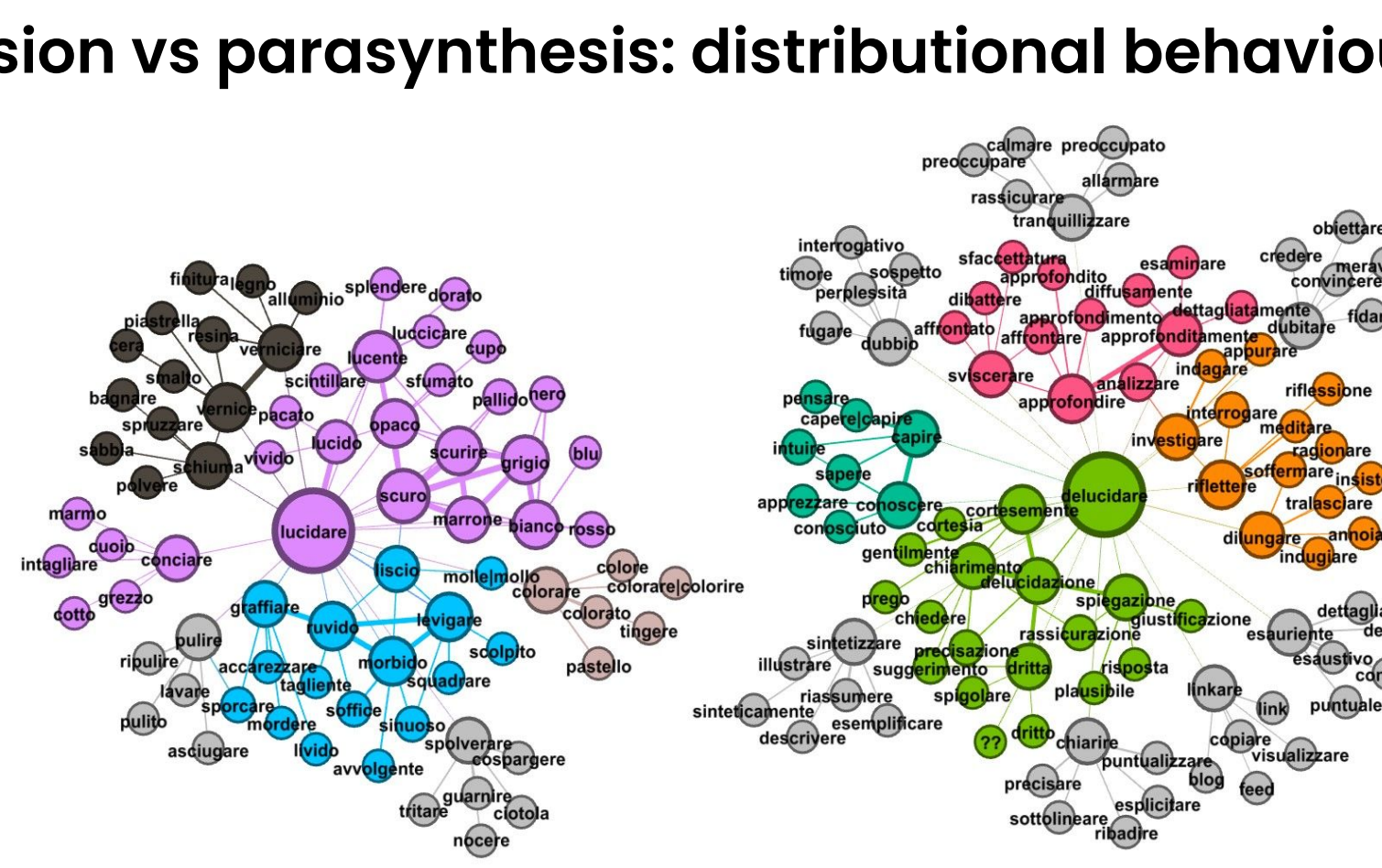
Two observations:

- There is a **significant difference** between the scores of the three groups, suggesting, on average, the **reliability of w2v scores** (despite a large standard deviation).
- Even though on average conversion and parasynthesis do not produce very similar pairs, there is **a group of highly semantically similar pairs** – among which, however, there is generally a high **difference in frequency** (median = 185.9%, calculated as absolute percentage difference)

c) Conversion vs parasynthesis: distributional behaviour of co-radicals

To find finer-grained differences between semantically similar co-radical verbs, we can plot their semantic graphs.

Semantic graphs were generated by extracting the **20 most similar neighbours of the two verbs**, as well as the 5 most similar neighbours of the 20 direct neighbours.



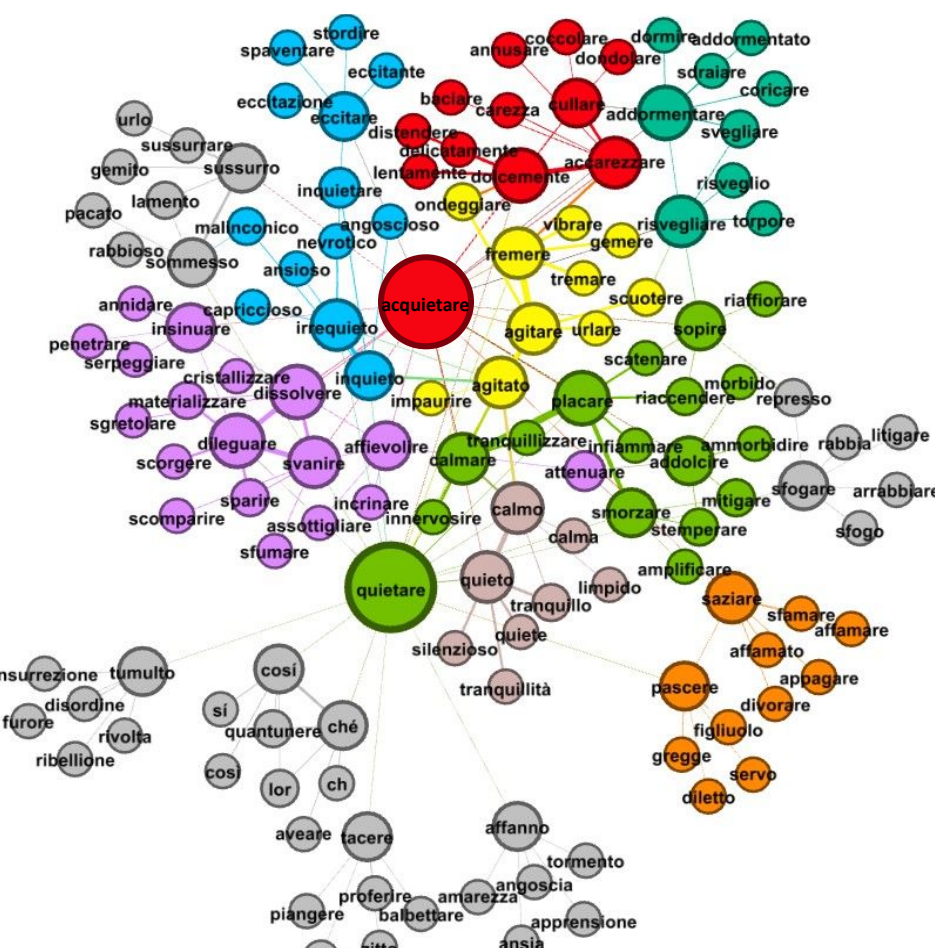
lucido 'shiny' 'lucid'

lucidare (conv.)
'to polish'
delucidare (paras.)
'to clarify'

similarity = 0.507414699
freq_difference = 167.72%

Not competing verbs:

1. low cosine similarity
2. distinct groups of neighbours



quieto 'calm'

quietare (conv.)
acquietare (paras.)
'to calm (down)'

similarity = 0.954316914
freq_difference = 7.59%

Competing verbs:

1. high cosine similarity
 2. share many neighbours
 3. very low frequency difference
- BUT**
1. At least one niche for *quietare*: not only psych states, but also the semantic field of *rebellion*, *insurrection*, etc.
 2. Hints of register variation: literary words closer to *quietare*, such as *pascere* 'to graze, to feed' and *figliuolo* 'son'.