Prediction of *to*-infinitives and gerunds as complements using Machine Learning

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## INDEX

- 1. Goal of the project
- 2. Methodology
- 3. First results
  - Extraction of features
  - ML Logistic regression
  - Decision trees
- 4. Discussions

## Goal of the project

DDD EXIMPLE SENTENCES

To predict the use of to-infinitives and gerunds as complements of a predefined set of verbs: Start doing vs. start to do

Lavin Lavin

	to-inf	-ing
like (in all forms)	11643	741
like (not in base form) <sup>1</sup>	2897	541
begin	18386	2617
continue	9742	911
hate	322	234
love	914	373
prefer	1947	96
propose	1157	191
start	6019	6666
attempt	6326	22
intend	6297	272
not bother	143	11

Table 1: Raw frequency in BNC (SketchEngine)

ENAMPLES

SENCT DON'S SENCT DENSONATIVE SENCE THUSANATIVE

## Methodology



#### **Extracting features from BNC:**

- Length of the non-finite verb
  - Begin to work
- Adjectives or nouns surrounding the non-finite verb
  - Wrong doing vs. To do wrong
- Argument structure of the non-finite verb.
  - Start to <u>eat</u>, start to <u>eat an apple</u>
- Temporal anchoring
  - Would like to do vs.?would like working

#### **Apply Machine Learning**





## First results: extraction of features

Two verbs **start** and **hate** and four features:

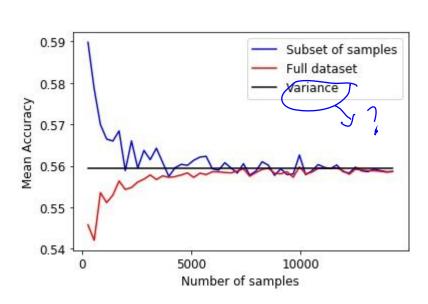


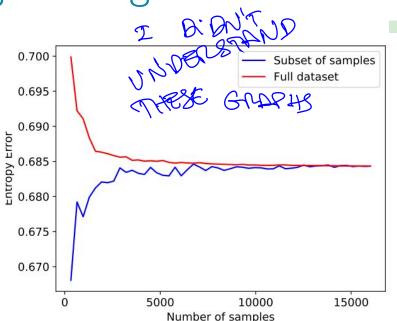
Feature	Values
Length	Number of characters
Adjs/N surrounding the verb	Before   After
Argument structure	No complement   Noun Phrase   Prep. Phrase
Tense	Present   Past   Future   Conditional
ATE BOLL 18.19	Present   Past   Future   Conditional  THE NOW - FINITE  AS FERNITE? (SEM ANTIC  LLASS))

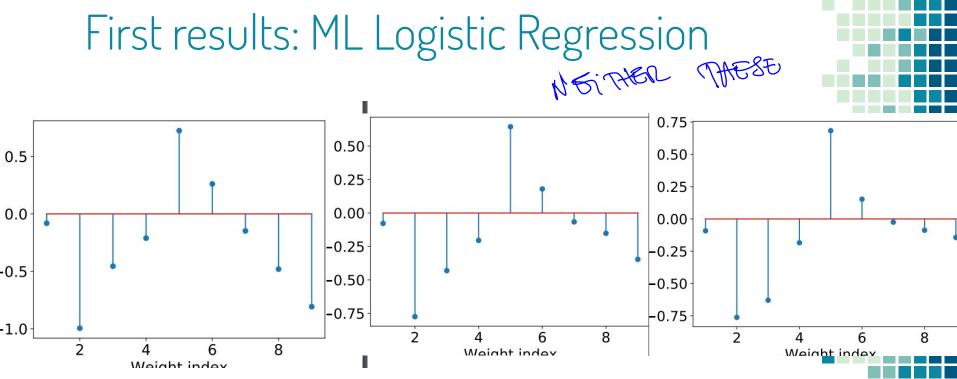
# First results: extraction of features

d	sent_id	finite_verb	non_finite	length	befaft_bef	befaft_aft	argstr_NC	argstr_N	P argstr_P	P tens	se_pre	tense_pst	tense_fut	tense_concverb_POS	target_form
3	2147	start	flow	4	0	0	0		0	1	0	C	1	0 VVG	1
4	3128	hate	kick	4	0	1	0		1	0	1	C	C	0 VVG	1
4	3262	start	write	5	0	0	1		0	0	1	C	C	0 VVG	1
5	4662	start	come	4	0	0	1		0	0	0	1	C	0 VV	0
5	4941	hate	hear	4	0	0	1		0	0	1	C	C	0 VV	0
6	5641	start	think	5	0	0	1		0	0	0	C	1	0 VV	0
6	6940	start	take	4	0	1	0		1	0	1	C	C	0 VV	0
6	7277	hate	see	3	0	0	0		0	1	0	1		0 VVG	1
6	7934	start	act	3	0	0	1		0	0	0	1	C	0 VVG	1
8	9543	start	transcribe	10	0	0	1		0	0	0	1	C	0 VV	0
8	10005	start	go	2	0	1	1		0	0	0	1	C	0 VV	0
8	10726	start	play	4	1	0	0		1	0	1	C	C	0 VV	0
8	10727	start	echo	4	0	0	0		0	1	1	C	C	0 VV	0
8	10730	start	distort	7	1	0	0		0	1	1	C	C	0 VV	0
8	10765	start	hear	4	0	1	0		1	0	1	C	C	0 VV	0_
8	11398	start	do	2	0	1	0		1	0	0	1	C	0 VV	0
8	11479	start	make	4	. 0	1	0		1	0	0	1	C	0 VV	0

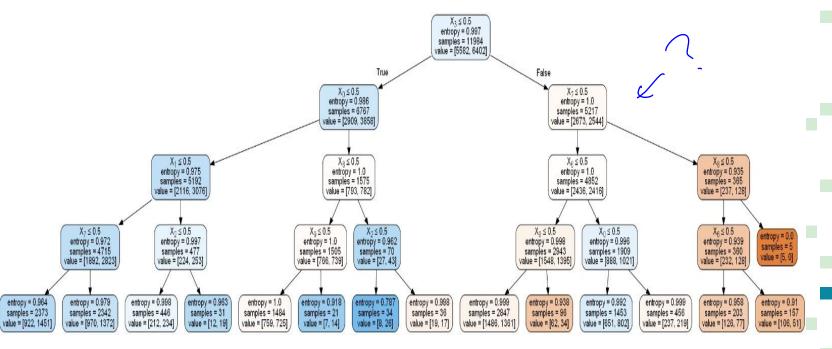
# First results: ML Logistic Regression



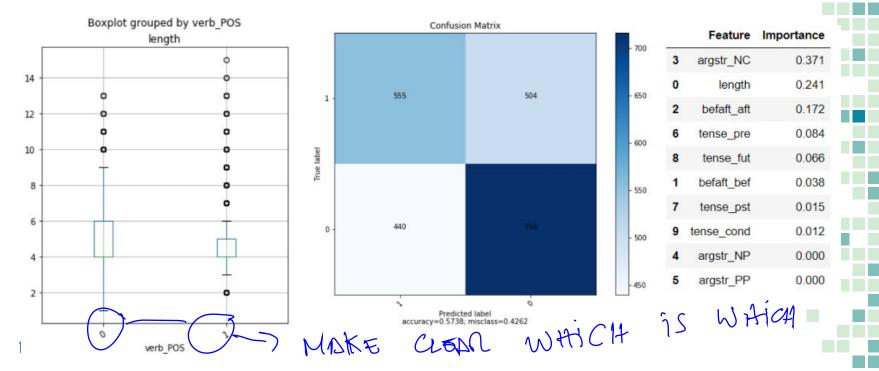




### First results: ML decision trees

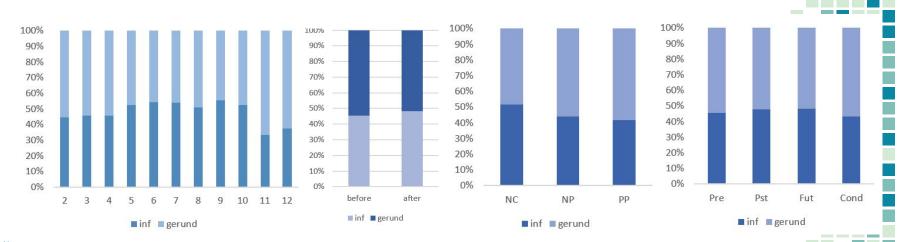


## First results: ML decision trees



#### Discussions

- Baseline: always predicting gerund, 0.53
- Difference between main verbs
  - Start and hate provide the most balanced data, other verbs tend to occur with infinitives
- Features may be improved:



## Further work

- Apply to more main verbs (begin, like, etc.)
- Extract more features:
  - Last mention of the non-finite verb
  - Animacy of the subject
  - Semantic class of main verb and non-finite verb
- Phonological features

  Try more ML methods

  Neural networks

  Mavbe. the two constructions are just Maybe, the two constructions are just

Interchangeable all the time.

JUTE NESTING. AT COULD BE THAT THEY AND,

STAGE! AND THAT'S WHY ONE GETS THE

STAGE! BUT ON IN THE CORPUS CENTINE.

CONLOBE CIPFERENT. POR INSTANCE, INTVITIVELY I'D AY Thank you! THAT "LNE TO X/ LOVE X-ING " ANE NOT Gràcies! USEP IN THE SQUE CON TEMS, Teşekkürler! ANOTHER ON BERION: WHY THE TO-CONSTIUCTION SO 谢谢! MUCH MORE PRERIENT