# Learning How to Conjugate the Romanian Verb Rules for Regular and Partially Irregular Verbs

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#### I. Stem alternations in Romanian verbs

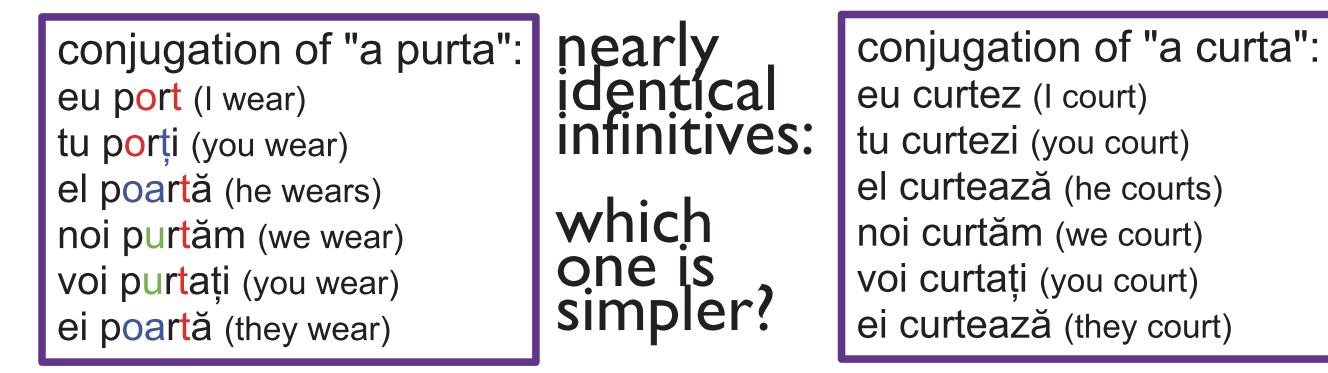
Stem alternations, or **apophony**, is one of the reasons why the Romanian language is difficult to acquire.

For **partially irregular verbs** it is not enough to learn a generic suffix variation pattern, because there are simultaneous variations in the stem.

#### II. Previous work

**Moisil** (1960): variable letters purta =  $pu_0 rt_0 a$  where:  $u_0 = \{u, oa, o\}, t_0 = \{t, t\}$ 

**Dinu, Ionescu** (2011, unpublished): context-sensitive rules to decode variable letters for some verbs. Idea: alternations are identifiable by their context.



# What is needed for automatic conjugation?

Romanian received a Latin-inspired classification of verbs into 4 conjugational classes, based on the ending of the infinitive form. This does not discriminate the two verbs shown above, so the standard model is **insufficient**.

The goal: given an infinitive form, know what letters change, and how they change. The trick: craft a sufficient, near-exhaustive, disjoint set of conjugation classes.

**Dinu et al** (2011): 7 conjugation classes for verbs ending in **-ta**. Knowing the class means knowing the alternations that occur. Idea: the classes can be learned.

A class corresponds to a **conjugation rule**: a set of 6 regular expressions matching the 6 conjugation forms of present tense verbs. Parts of the forms that are not accounted for must remain fixed, i.e. a rule accounts for all the variation.

Classification using **character n-gram** features + SVM: n-gram size chosen to be 3 for model simplicity (n  $\sim$ = 5 is optimal) Input: 'purta' => 'p', 'u', 'r', 't', 'a', 'pu', 'ur', 'rt', 'ta', 'pur', 'urt', 'rta' Output: label in {1, 2, 3, 4, 5, 6, 7}, imbalanced classes Results: 82.71% accuracy and 80% F-score

### III. Crafting conjugation rules using regular expressions

#### Process

To manually expand a set of conjugation rules:

. Select unmatched verb 2. Add rule to completely conjugate it 3. Match verbs against new rules

#### For example: I. Say the verb 'a omorî' (to kill) is not matched 2. A conjugation rule matching this verb would be:

#### **Results**:

We threw out rules covering <4 cases, leaving 30 rules covering 95% of the verbs

# Rules overview:

rule #: size: rule #: size: 547 16 13

#### Interaction between rules

The largest covering rule has no alternations in the root, just the suffix. Other rules model 0-2 apophonys. Some rules correspond to the same variable letter, but it varies differently. For example:

# Some rules overlap:

	rule 10 (a cânta)	rule 12 (a deștepta)	rule 13 (a deșerta)	rule 15 (a desfăta)
1sg	^(.*)t\$	^(.*)e(.*)t\$	^(.*)e(.*)t\$	^(.*)ăt\$
	^(.*)ți\$	^(.*)e(.*)ti\$	^(.*)e(.*)ti\$	^(.*)eti\$
3sg	^(.*)tă\$	^(.*)ea(.*)tă\$	^(.*)a(.*)tă\$	^(.*)ată\$
1pl	^(.*)tăm\$	^(.*)e(.*)tăm\$	^(.*)e(.*)tăm\$	^(.*)ătăm\$
2pl	^(.*)tați\$	^(.*)e(.*)tați\$	^(.*)e(.*)tați\$	^(.*)ătați\$
3pl	^(.*)tă\$	^(.*)ea(.*)tă\$	^(.*)a(.*)tă\$	^(.*)ată\$

l sg:	^(.*)0(.*)\$	omor
2sg:	^(.*)o(.*)i\$	omori
3sg:	^(.*)oa(.*)ă\$	omoară
l pl:	^(.*)o(.*)âm\$	omorâm
2pl:	^(.*)o(.*)âți\$	omorâți
3pl:	^(.*)oa(.*)ă\$	omoară

3. This rule also matches, among others, the verb 'a doborî' (to defeat) so mark this one as matched too.

2	8	17	6
3	18	18	4
4	5	19	14
5	8	20	124
6	16	21	25
7	3330	22	15
8	273	23	7
9	89	24	41
10	4	25	51
11	5	26	185
12	4	27	1554
13	106	28	486
14	13	29	5
15	5	30	27

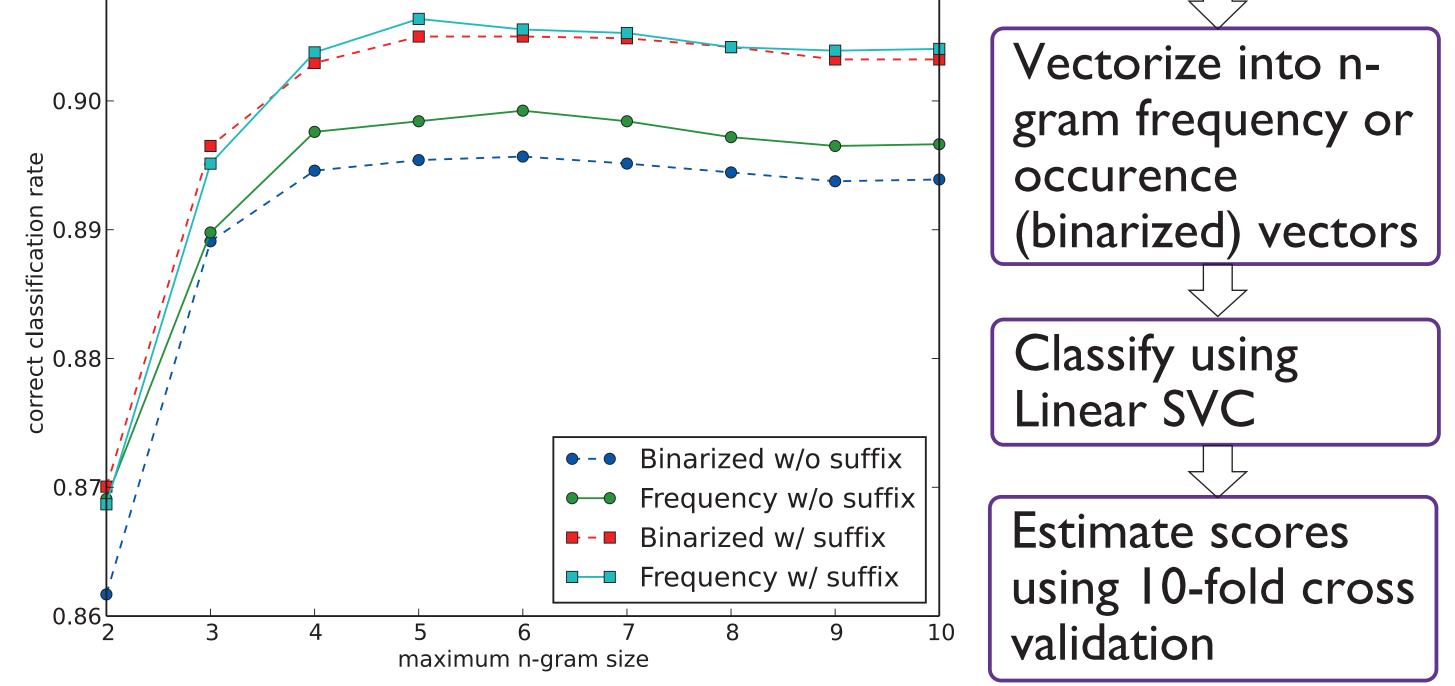
Rule 13 is much more productive than 10, 12 and 15, but we miss the importance of the t-t alternation itself. It also occurs in rule 14, 21 and also verbs with too rare conjugation patterns to generalize ('a purta' is actually a singleton!)

#### IV. Classification methodology Add suffix marker Extract Dataset of 0 'purta' -> 'purta\$' indicative Romanian verb present tense, forms label infinitives Extract n-grams up **Results**: to size n 0.91

# V. Conclusion and perspectives

# Estimated scores:

Parameters chosen by grid search: n=5, append '\$', do not binarize, C=0.1 Correct classification rate: 90.64% (baseline choosing most probable class: 48%) Weighted averaged precision: 80.90%, recall: 90.64%, F<sub>1</sub> score: 89.89%. Appending the artificial terminator marker '\$' consistently improves accuracy by around 0.7% irrelevant of the other parameters. Frequency features perform slightly better than binarized ones for this task



#### What does this mean?

Verb conjugation can be learned with good scores, even with the assumption that classes don't interact. Our classes are **coarse-grained**. An **exhaustive model**, at least for the training data, will need to have many classes for unique and near-unique conjugation patterns. For better generalization: we need a finer-grained system.

## Future work and collaboration ideas:

Build a more compact model by eliminating rule interaction: (see discussion above) Compare with hand-crafted rule based conjugation Try human evaluation on unseen, unlabeled verbs Actually build a verb conjugation using classification output (trivial) Extend to other languages with similar behaviour (Hebrew)