

Agent noun formation in Czech: An empirical study on suffix rivalry

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Introduction: Agent nouns across languages

- one of the most frequent categories attested cross-linguistically (Bauer 2002, Štekauer et al. 2012)
- derived from verbs (nomina agentis)
 - *writer* < *write*
- agentive meaning ascribed also to denominal nouns (Rainer 2015; nomina actoris)
 - *paintballer* < *paintball*
- often both a directly related noun and verb attested (oed.com):
 - *fisher* < *fish.v* (*fish.v* < *fish.n*)
 - *footballer* < *football.n* or *footballer* < *football.v* (*football.v* < *football.n*)

Agent noun formation in Czech

- 35 different agent suffixes to combine with verbs (Daneš et al. 1967, Dokulil et al. 1986, Štícha et al. 2018)

- 8 most frequent of them covered by the paper:

- | | |
|---|--|
| a. <i>uč-i-tel</i> ‘teacher’ < <i>uč-i-t</i> ‘to teach’ | e. <i>soud-ce</i> ‘judge’ < <i>soud-i-t</i> ‘to judge’ |
| b. <i>řid-i-č</i> ‘driver’ < <i>říd-i-t</i> ‘to drive’ | f. <i>kuř-ák</i> ‘smoker’ < <i>kouř-i-t</i> ‘to smoke’ |
| c. <i>řez-ník</i> ‘butcher’ < <i>řez-a-t</i> ‘to cut’ | g. <i>kup-ec</i> ‘buyer’ < <i>koup-i-t</i> ‘to buy’ |
| d. <i>kov-ář</i> ‘blacksmith’ < <i>kov-a-t</i> ‘to forge’ | h. <i>mluv-čí</i> ‘speaker’ < <i>mluv-i-t</i> ‘to speak’ |

- *-tel* only in agents, but most of the suffixes convey more than one semantic category:

e.g. the suffix *-ec* in

1. agents (*letec* ‘pilot’ < *létat* ‘to fly’),
2. inhabitants (*Nepálec* ‘Nepali’ < *Nepál* ‘Nepal’),
3. bearers of social roles (*vdovec* ‘widower’ < *vdova* ‘widow’),
4. bearers of qualities (*stařec* ‘old man’ < *starý* ‘old’),
5. animal names (*dravec* ‘predator’ < *dravý* ‘predatory’),
6. instruments (*bodec* ‘spike’ < *bodat* ‘to stab’),
7. toponyms (*Hradec* < *hrad* ‘castle’), etc.

1. Design of the data
 - A data-based approach to the agent suffix rivalry
 - Extraction of the agent nouns from the corpus
 - Features to assign
2. Baseline solution
3. Machine learning experiments: logistic regression vs. decision trees
 - Experiments on all features
 - Experimenting with feature sets
4. Discussion & conclusions
 - Comparison of the methods
 - Incorrect predictions
 - Final remarks

A data-based approach to the agent suffix rivalry

- paradigmatic approach (Bonami & Strnadová 2019)
 - agent nouns as members of morphological families
 - all potential predecessors considered

| agent noun | verb.IPVF PFV | noun | adjective |
|----------------------------------|---|--------------------------|------------------------------|
| <i>sjednot-i-tel</i> ‘unifier’ | - <i>sjednot-i-t</i> ‘unify’ | | |
| <i>sjednoc-ova-tel</i> ‘unifier’ | <i>sjednoc-ova-t</i> - ‘unify’ | | |
| <i>model-ář</i> ‘modeler’ | <i>model-ova-t</i> - ‘model’ | <i>model</i> ‘model’ | |
| <i>zvon-ík</i> ‘bell-ringer’ | <i>zvon-i-t</i> - ‘ring’ | <i>zvon</i> ‘bell’ | |
| <i>závod/n/ík</i> ‘racer’ | <i>závod-i-t</i> - ‘race’ | <i>závod</i> ‘race’ | <i>závod-n-í</i> ‘racing’ |
| <i>boj-ov/n/ík</i> ‘fighter’ | <i>boj-ova-t</i> - ‘fight’ | | <i>boj-ov-n-ý</i> ‘fighting’ |
| <i>střel-ec</i> ‘shooter’ | <i>stříl-e-t</i> <i>střel-i-t</i> ‘shoot’ | <i>střel-a</i> ‘shot’ | |
| <i>kup-ec</i> ‘purchaser’ | <i>kup-ova-t</i> <i>koup-i-t</i> ‘purchase’ | <i>koup-ě</i> ‘purchase’ | |

Extraction of the agent nouns from the corpus

- all masculine animate nouns ending in one of the suffix strings extracted from the SYN2015 corpus (Křen et al. 2015)
- non-agents, nouns where the string is not a suffix, compounds, typos, etc. excluded
- potential predecessors listed: verb (imperfective | perfective), noun, adjective
- nouns without a verbal predecessor removed

>>> 1,178 nouns in the final set

| Suffix | -tel | -č | -ník/-ík | -ář/-ař | -ce | -ák | -ec | -čí | Σ |
|--------|------|-----|----------|---------|-----|-----|-----|-----|--------------|
| Count | 426 | 388 | 106 | 96 | 66 | 50 | 32 | 14 | 1,178 |

- 20 features assumed as potentially relevant for modeling the rivalry (Strnadová 2015, Santana-Lario & Valera 2017, Bonami & Thuilier 2019, Wauquier et al. 2020)

Features to assign

- related to the motivating verb(s)
 - final consonant of the root
 - number of prefixes
 - theme
 - aspect
 - conjugation class
- related to the derivational paradigm
 - which motivating items available?
 - does the verb have a suffixed aspectual counterpart?
 - does an inanimate homonym exist?
 - absolute corpus frequency of all items
 - motivating items ordered by frequency

válečník *válčit* – *válka* – *válečný*
 warrior make war – war.n – war.adj

| | |
|---------------------------|-----------------|
| target_noun_suffix | -ník -ík |
| root_final | č |
| root_final_cvs | consonant |
| root_final_vertical | fricative |
| root_final_horizontal | postalveolar |
| number_prefixes | 0 |
| v1_theme | i |
| v1_aspect | imp |
| v1_conjug | 4 |
| v1_suf_asp_counterpart | no |
| v2_theme | – |
| v2_aspect | – |
| v2_conjug | – |
| paradigm_type | NNA-V- |
| inanim_noun | no |
| freq_parent_noun | 25,895 |
| freq_parent_adj | 4,953 |
| freq_parent_v1 | 499 |
| freq_parent_v2 | – |
| freq_slots | VAN |

Baseline solution

- data set divided into a training set, an evaluation set, and a hold-out set (60:20:20)
- random baseline predicting one of the eight suffixes in a uniform distribution
 - weighted average of **F-score=0.16** calculated on the hold-out data set

| Suffix | all | -tel | -č | -ník/-ík | -ář/-ař | -ce | -ák | -ec | -čí |
|-----------|-------------|------|------|----------|---------|------|------|------|------|
| Instances | 233 | 85 | 77 | 21 | 19 | 13 | 10 | 6 | 2 |
| Precision | 0.28 | 0.43 | 0.32 | 0.10 | 0.07 | 0.08 | 0.04 | 0.04 | 0.04 |
| Recall | 0.13 | 0.14 | 0.10 | 0.14 | 0.11 | 0.23 | 0.10 | 0.17 | 0.50 |
| F-score | 0.16 | 0.21 | 0.16 | 0.12 | 0.09 | 0.12 | 0.05 | 0.06 | 0.07 |

Machine learning experiments

- which agent suffix is chosen by a particular verb?
 - the agent suffix used as the target class in the experiments
 - the other features as predictors
- two different machine learning methods applied
 - hyper-parameter settings tuned in the first experiment on all features
 - results compared to experiments on four different feature subsets

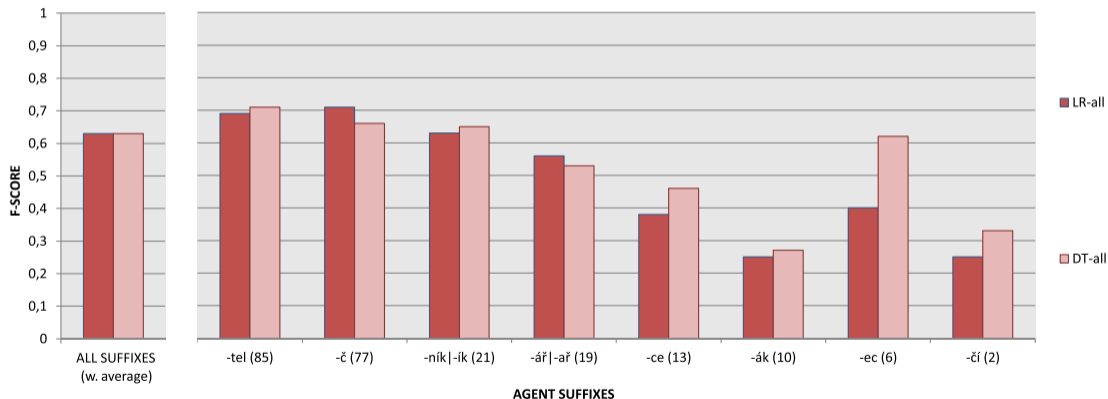
- Logistic regression

```
classifier_LR = LogisticRegression(  
    multi_class='multinomial',  
    class_weight='balanced',  
    solver='newton-cg',  
    penalty='l2',  
    C=1e30)
```

- Decision trees

```
classifier = DecisionTreeClassifier(  
    criterion='entropy',  
    class_weight='balanced',  
    splitter='best',  
    max_depth=10)
```

Experimenting with all features: F-score on hold-out data

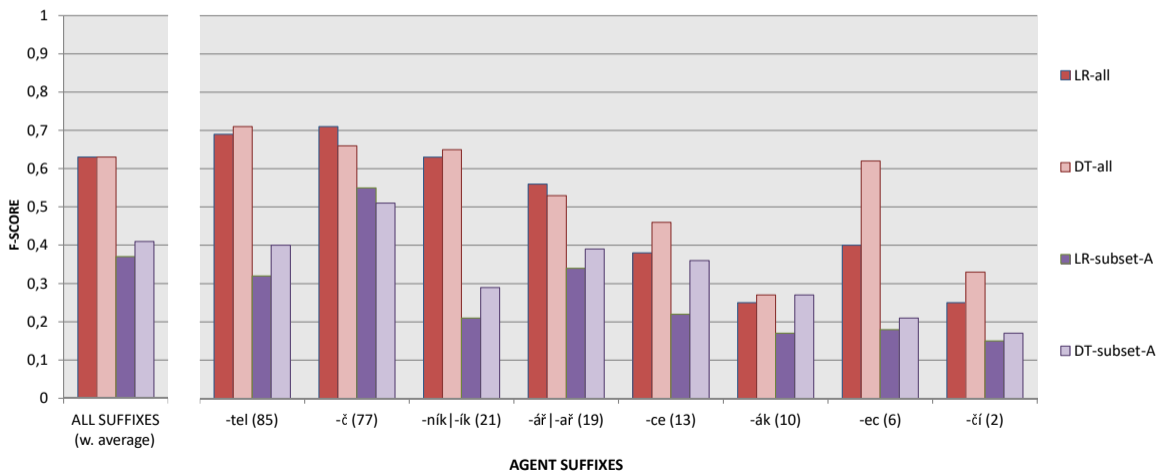


Experimenting with feature subsets: Subsets A to D

- A: the motivating verb(s): root's final character and theme
[root_final, root_final_cvs, root_final_vertical, root_final_horizontal, v1_theme, v2_theme]
- B: the motivating verb(s): number of prefixes, theme, aspect, conjugation class
[number_prefixes, v1_theme, v1_aspect, v1_conjug, v2_theme, v2_aspect, v2_conjug]
- C: the derivational paradigm: which motivating items available?, does the verb have a suffixed aspectual counterpart?, does an inanimate homonym exist?
[paradigm_type, v1_suf_asp_counterpart, inanim_noun]
- D: corpus frequency of the motivating items
[freq_parent_noun, freq_parent_adj, freq_parent_v1, freq_parent_v2, freq_slots]

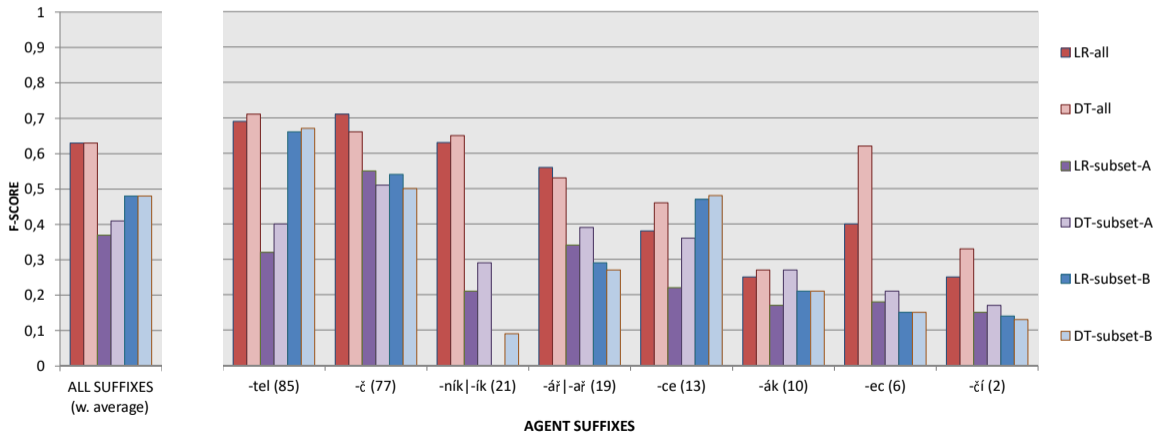
Experiments with the subset A: F-score on hold-out data

subset A: root_final, root_final_cvs, root_final_vertical, root_final_horizontal, v1_theme, v2_theme



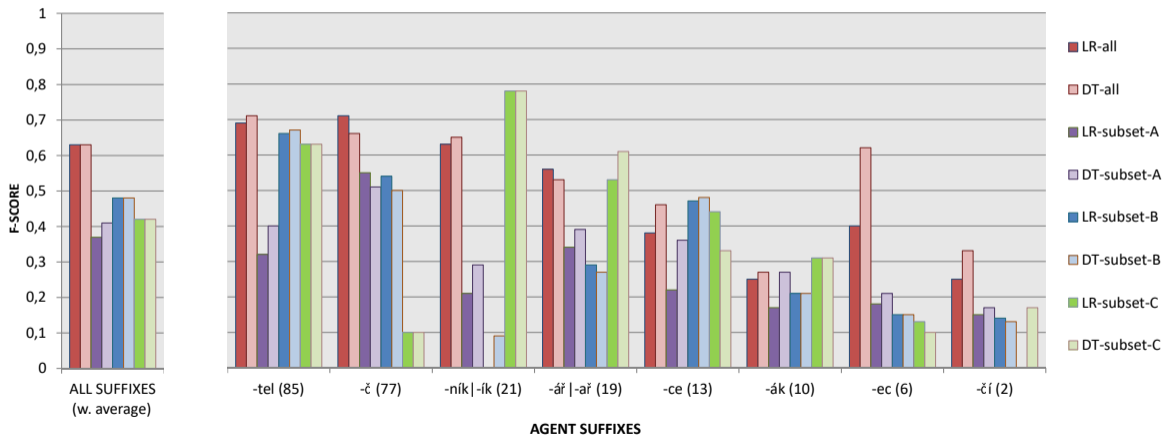
Experiments with the subset B: F-score on hold-out data

subset B: number_prefixes, v1_theme, v1_aspect, v1_conjug, v2_theme, v2_aspect, v2_conjug



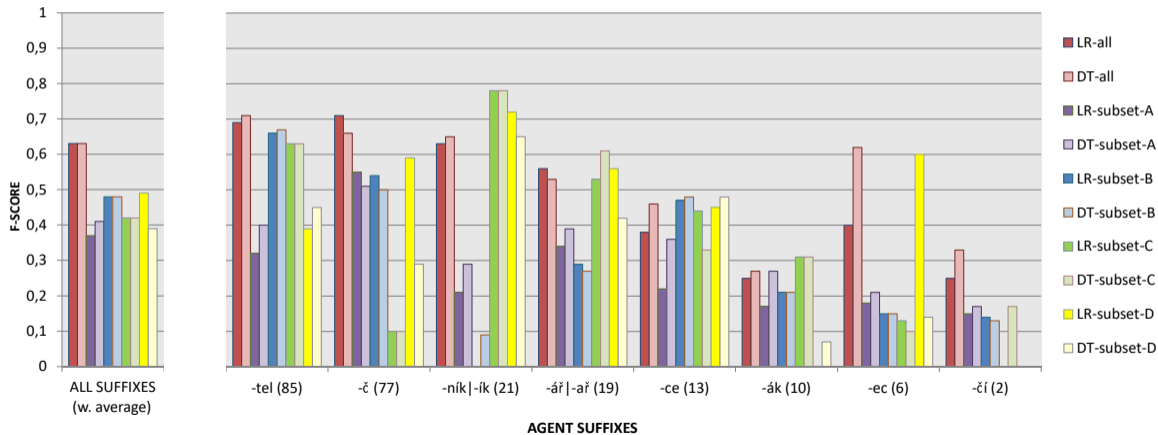
Experiments with the subset C: F-score on hold-out data

subset C: paradigm_type, v1_suf_asp_counterpart, inanim_noun



Experiments with the subset D: F-score on hold-out data

subset D: freq_parent_noun, freq_parent_adj, freq_parent_v1, freq_parent_v2, freq_slots



Discussion: predicting all suffixes by logistic regression vs. decision trees

- the methods model the impact of the features differently
 - logistic regression estimates dependencies among the given features
 - decision trees propose a set of decisions over the features such that their disorder (entropy) is minimized
- all suffixes best predicted based on all features
 - logistic regression with all features: F-score=0.63
 - decision trees with all features: F-score=0.63 (vs. baseline F-score=0.16)
- features seem to be relevant
- there must be more relevant features not yet covered by the data

Results on individual suffixes

- *-tel*, *-č*, *-ec*, *-čí*: best results with all features
- *-ce* the same results on the subset B (detailed features of the verb) and D (frequency)
- *-ník|-ík*, *-ář|-ař*, *-ák* best predicted from the derivational paradigm (subset C)
 - *-ník|-ík* motivated by a verb/verbs and by an adjective (*pracovník* ‘worker’)
 - *-ář|-ař* motivated by a noun and a verb/verbs, never has an inanimate homonym (*záchránář* ‘rescuer’, *tiskař* ‘printer’)
 - *-ák* based on a verb/verbs, can have an inanimate homonym (*piják* ‘drunkard x blotter’)
- subset A (root & themes) not sufficient

| suffix | noun | all features (log.regr./dec.trees) | A | B | C | D |
|-----------------|------|---------------------------------------|-----------|-----------|-----------|-----------|
| <i>-tel</i> | 85 | 0.69/0.71 | 0.32/0.40 | 0.66/0.67 | 0.63/0.63 | 0.39/0.45 |
| <i>-č</i> | 77 | 0.71/0.66 | 0.55/0.51 | 0.54/0.50 | 0.10/0.10 | 0.59/0.29 |
| <i>-ník -ík</i> | 21 | 0.63/0.65 | 0.21/0.29 | 0.00/0.09 | 0.78/0.78 | 0.72/0.65 |
| <i>-ář -ař</i> | 19 | 0.56/0.53 | 0.34/0.39 | 0.29/0.27 | 0.53/0.61 | 0.56/0.42 |
| <i>-ce</i> | 13 | 0.38/0.46 | 0.22/0.36 | 0.47/0.48 | 0.44/0.33 | 0.45/0.48 |
| <i>-ák</i> | 10 | 0.25/0.27 | 0.17/0.27 | 0.21/0.21 | 0.31/0.31 | 0.00/0.07 |
| <i>-ec</i> | 6 | 0.40/0.62 | 0.18/0.21 | 0.15/0.15 | 0.13/0.10 | 0.60/0.14 |
| <i>-čí</i> | 2 | 0.25/0.33 | 0.15/0.17 | 0.14/0.13 | 0.00/0.17 | 0.00/0.00 |
| <i>all</i> | 233 | 0.63/0.63 | 0.37/0.41 | 0.48/0.48 | 0.42/0.42 | 0.49/0.39 |

Incorrect predictions

- -ník/ík predicted in **signatník* (expected *signatář* ‘signatory’)
 - the native suffix incompatible with the foreign base (cf. German *Signatar*)
- -č predicted in **oblehač* (vs. *oblehatel* ‘besieger’), **budič* (vs. *buditel* ‘revivalist’)
 - differences in registers (formal register of the base vs. informal suffix)
 - *budič* attested as an inanimate noun
- -ce predicted in **ulejvce* (vs. *ulejvák* ‘loafer’), **výčepce* (vs. *výčepák* ‘bartender’)
 - different registers (informal base vs. formal suffix)

Conclusions

- study on rivalry among eight suffixes used in Czech agent nouns
- 1,178 agent nouns with verbal predecessors
 - provided with 20 features (phonology, morphology, paradigmatic info)
- random baseline model's F-score 0.16
- two machine-learning methods applied
 - experiments with all features vs. with feature subsets
 - best prediction of all suffixes based on all features
 - F-score 0.63 both with logistic regression and decision trees
 - derivational paradigms relevant for predicting individual suffixes
- not considered:
 - diachronic features (date of attestation), registers, origin (foreign vs. native)
 - speakers's preferences, lexicalization

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