

Generalizing French schwa deletion: the role of indexed constraints

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Intro

- Two possible problems for learned models:
 - Not capturing training data (underfitting)
 - Not generalizing from training data (overfitting)
 - Tendency: tradeoff between these (Hastie et al. 2001)
- Indexed constraints designed to capture training data (Pater 2000)
 - Will this hurt generalization?
- Tested with 4 procedures for indexed constraint discovery
 - Learn from French schwa deletion corpus
 - Apply learned models to existing experimental data

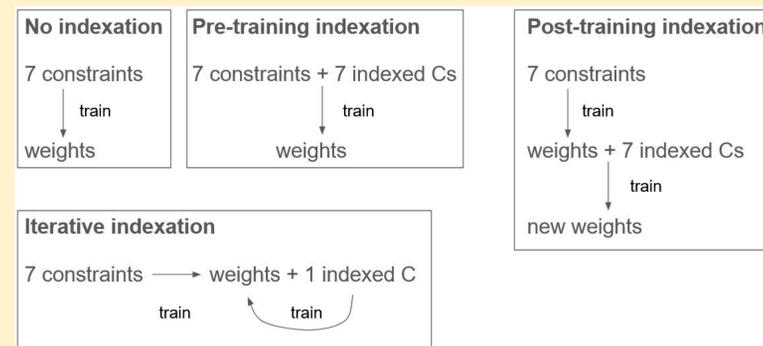
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Procedure

- MaxEnt (Goldwater & Johnson 2003)
 - batch-trained using hgR (Staubs 2011) from all-0 weights
- Start out without indexed constraints
 - Use word-specific gradients of constraint weights to find indexed constraints (cf. Becker 2009, Pater 2010)
- 4 increasingly involved indexation procedures
 - No indexation
 - Pre-training (use weight gradients before training)
 - Post-training (use weight gradients after training once)
 - Iterative (add one indexed constraint at a time while training in between, until convergence)



French schwa deletion (Dell 1985)

- Deletion of /ə/ (phonetically [œ]) optional in many contexts
- Rate of deletion depends on phonological context
- However, rate of deletion also depends on individual word

e.g., VC_VC > #C_C > CC_C, C_CC
 kas(ə)ʁɔl s(ə)ʁɛ̃ sub(ə)so
 'pot' 'canary' 'jolt'

Examples from
Racine (2008) corpus

e.g. /səmə̃/ 'week' (50% deletion)
 /səmə̃stɛ/ 'semester' (14% deletion)

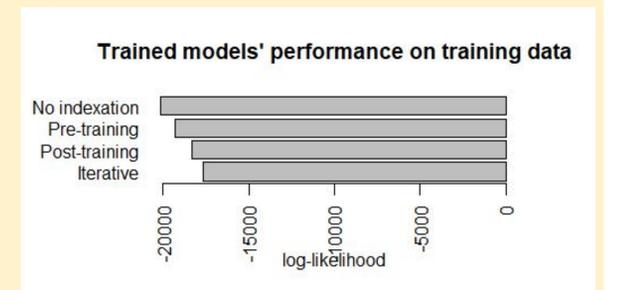
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Training on corpus

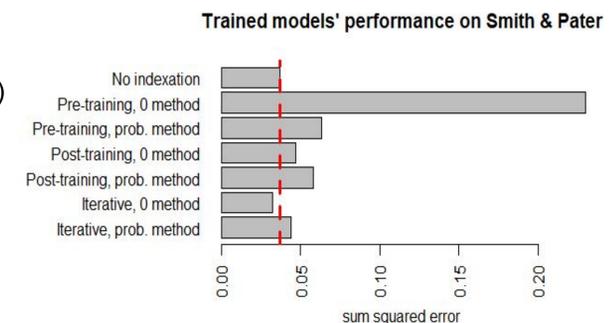
- Corpus: 456 words × 12 France French speakers (Racine 2008)
 - Judgments on schwa-ful and schwa-less variants of same word
 - transformed into pseudo-frequencies
- Constraints:
 - *ə, *ə/non-penult, Max, *#CC, *CCC, *CNC, *CTN (cf. Kaplan 2011)
- Evaluation: log-likelihood of training data (closer to 0 = better)
- As expected: data captured better as indexation gets more involved
- Performance not tied to number of indexed constraints:
 - Pre-training: 7
 - Post-training: 7
 - Iterative: 4

T = plosive
N = nasal



Generalization to Smith & Pater (2020)

- Smith & Pater (2020): Participants choose between schwa or no schwa in French phrases containing -CCə and -VCə words followed by CV or CVCV
- Use trained procedures to predict proportion of schwa responses per context
 - Constraints: *ə, *ə/non-penult, Max, *CCC, and any indexed versions
 - Generalization to experimental data:
 - 0 method: no indexed constraint violations
 - probabilistic method: indexed constraint violations determined by Maximum Likelihood (cf. Becker 2009)
- Evaluation: sum squared error (SSE; smaller = better)
 - More involved indexation = better
 - 0 method of generalization slightly better?
 - Except for pre-training indexation: definitely worse
 - Best indexation procedure on par with No indexation



Conclusions

- Indexed constraints improve account of training data, but do not have to hurt generalization!
 - Similar to how adding random effects improves generalization in Mixed Effects models (e.g., Zymet 2018; Barr et al. 2013)
- Best generalization behaviour when indexed constraints induced one by one, but not used in generalization to new items
 - Non-indexed constraint weights improved by precense of indexed constraints
- When evaluating on new data, method of generalization matters
 - For example, pre-training procedure fares much worse with 0 method