

Generalizations and conspiracies

Edward Stabler (UCLA)
Kristine M. Yu (UMass Amherst)

MG+1 2023

Some ideas emerging from Stabler&Yu 2023, etc,
on specifying and implementing MGs and interfaces

these slides: <https://github.com/epstabler/star>

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How often? (https://github.com/epstabler/star)

...when faced with different concerns, we should separate
them as completely as possible, and deal with them in turn.

EWD636

You see, while we all know that unmastered
complexity is at the root of the misery, we
do not know what degree of simplicity
can be obtained, nor to what extent
the intrinsic complexity of the whole
design has to show up in the interfaces.

EWD1304-2

Taming the complexity

- 0.1 MG syntax + OT prosody (Stabler&Yu'23)
- 0.2 MG syntax with agreement (Ermolaeva&Kobele'23)
- 0.3 MG syntax + logical transduction (Vu,De Santo&Dolotion'22)
- 0.4 Incremental comprehension, production (Hunter&al'19)
- 0.5 MG syntax with head movement (Stanojevic'17)
- ...

- all these are difficult to understand, combinations more so!

⇒ **Factoring: Discover (additional) modularity**

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Taming the complexity

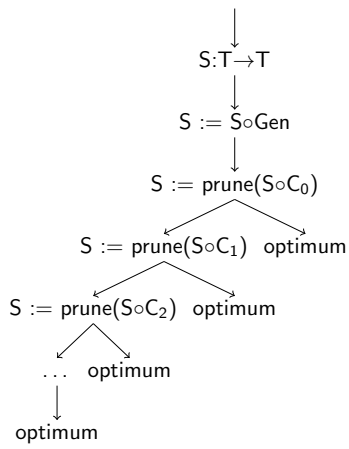
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When talking to Chomsky and Chomskians:
concern that MGs do not recognize and pull out
internal-external merge similarities, labeling, linearization.

They are right.
There have been some proposals, but best way has been unclear.
Making these basics easy will reveal/enable further extensions.

Bennett&al Irish prosody



syntax

gen candidates

constrain, prune until deterministic

NB: transducer compositions at every step

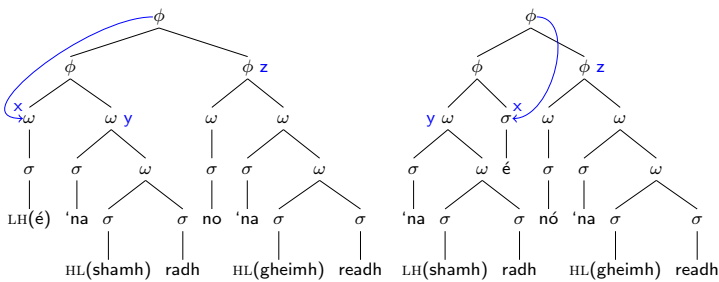
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Bennett&al Irish prosody

Bennett&al propose that an optimal prosodic structure is determined not just by syntax, but by constraints sensitive to phonological features like stress and accent. And in some cases, the optimal prosodic structure is even one that modifies linear order.

Stabler&Yu capture the Bennett&al analysis like this: Initially, syntax is composed from parts, but here we can think of it as an identity transduction on a single tree, from which gen generates possible candidates, including some with modified linear order; then optimal structure selected by evaluating constraints in rank order

The operations here apply not to particular structures, but to finite transducers, each generating possibly infinitely many structures



Transduction 'lowers' if **x prosodically weak** (≠syntax!)

predP(DP(x),PredP(y,z)) → $\phi(\phi(x,y),z)$

predP(DP(x),PredP(y,z)) → $\phi(\phi(y,x),z)$

(Here: • 'unextend' rules, • get rid of parent-child, • sharing)

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Note!! • the trigger is not syntactic, • the effect is not one syntax could achieve, and • that effect has no semantic consequences. I.E. This is not syntax!

The SCiL paper/talk gets to here, but lacks space expose ingredients in full generality.

Unextend for transducer composition

predP(DP(q0(x),PredP(q1(y),q2(z)))) → q3(φ(φ(x, y), z)))

Engelfriet&al: For composition, compile to ‘1-normal form’:

DP(q0(x)) → q3(x)
PredP(q1(y),q2(z)) → q4(y, z)
predP(q3(x),q4(y, z)) → q5(x, y, z)
q5(x, y, z) → q6(φ(x, y), z)
q6(x, y) → q3(φ(x, y))

... Then compositions are fast, but complexity multiplies
1-normal forms and compositions aren’t for human reading!
They conspire to exactly implement the generalization.

Factoring the tree transductions

Standard tree automata not quite what we need, so dissect a bit. . .

Tree automata recursively ‘fold’ classification, construction steps through tree traversal (e.g. Bahr’12)

- label: Feats⁺ → Feats
- fold: accumulate changes while traversing structure
(a → b → b) → b → t a → b

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Unextend for transducer composition

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At SCiL, we left the states out of the rules, added back in red.
This 1 rule has 3 input symbols, 2 output symbols, and 4 states.
(Not too complicated – syn/prosody research can work at this level)

But we implement that rule with a set of simpler rules.
To handle complex transducer i/o: 1-normal form.

Then to compose, intuitively: when first transducer outputs a symbol, it can immediately be given to second transducer.
(multiplies rules. . . implementation-oriented, only for psycholinguists & al)

NB: No one of these 1-normal rules can be justified independently of the others. It is truly the extended rule (= the whole conspiracy of 1-normal rules) that is supported by the evidence.

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Refactor and adjust 1: *-extensions

- Multiple successors: left, right, etc?
But then how to express unranked generalizations?
- Better: replace (parent,child) relation with (parent, next gen) where next gen is unbounded.
E.g. data Tree a = Node a [Tree a],
with regular *-sequence [Tree a].
- OK, but some components still too complex:
esp. labeling with k 'movers' to emulate sharing

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With *-extension, the parent-child (successor) relation is no longer basic

Refactor and adjust 2: DAG transducers

Labeling simplified when sharing is real: not trees, DAGs

⇒ Split structure up, for arity k labels (Fowlie'11)

Fold generalizes to DAGs (Priese'07, Bahr&Axelsson'17)
E.g. data Dag a = {(Node a, [Edges])}

DAG automata recursively 'fold' classification, construction steps through graph traversal

Weighted search methods extend to *-DAG grammars (Höfner&Möller'12, Kidney&Wu'21, Gibbons&al'22)

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In progress

- **Appropriate modularity over structures with empirical support**
 - decreases specification complexity (better theory!), and
 - exposes intermodular dependencies, some of which fold through structures recursively,
 - suggests specialized automata for psych modeling
- High-level generalizations → lower-level conspiracies: multiple processes, different granularities, in concert
- Tree/dag transducer composition 'platform', in prep – a library with functions for basic fold, labeling, composition ops

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Conspiracies happen, but not in the theoretical statement, which is optimized for simplicity, intelligibility, and clear relevance to evidential support/learnability

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