

Efficient Generation in Primitive Optimality Theory

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Abstract

This paper introduces primitive Optimality Theory (OTP), a linguistically motivated formalization of OT. OTP specifies the class of autosegmental representations, the universal generator **Gen**, and the two simple families of permissible constraints. In contrast to less restricted theories using Generalized Alignment, OTP's optimal surface forms can be generated with finite-state methods adapted from (Ellison, 1994). Unfortunately these methods take time exponential on the size of the grammar. Indeed the generation problem is shown NP-hard in this sense. However, techniques are discussed for making Ellison's approach fast in the typical case, including a simple trick that alone provides a 100-fold speedup on a grammar fragment of moderate size. One avenue for future improvements is a new finite-state notion, "factored automata," where regular languages are represented compactly via formal intersections $\cap_{i=1}^k A_i$ of FSAs.

1 Why formalize OT?

Phonology has recently undergone a paradigm shift. Since the seminal work of (Prince & Smolensky, 1993), phonologists have published literally hundreds of analyses in the new constraint-based framework of Optimality Theory, or OT. Old-style derivational analyses have all but vanished from the linguistics conferences.

The price of this creative ferment has been a certain lack of rigor. The claim for OT as Universal Grammar is not substantive or falsifiable without formal definitions of the putative Universal Grammar objects **Repns**, **Con**, and **Gen** (see below). Formalizing OT is necessary not only to flesh it out as a linguistic theory, but also for the sake of computational phonology. Without knowing what classes of constraints may appear in grammars, we can say only so much about the properties of the system,

or about algorithms for generation, comprehension, and learning.

The central claim of OT is that the phonology of any language can be naturally described as *successive filtering*. In OT, a phonological grammar for a language consists of a vector C_1, C_2, \dots, C_n of soft **constraints** drawn from a universal fixed set **Con**. Each constraint in the vector is a function that scores possible output representations (surface forms):

$$(1) C_i : \mathbf{Repns} \rightarrow \{0, 1, 2, \dots\} \quad (C_i \in \mathbf{Con})$$

If $C_i(R) = 0$, the output representation R is said to **satisfy** the i th constraint of the language. Otherwise it is said to **violate** that constraint, where the value of $C_i(R)$ specifies the degree of violation. Each constraint yields a filter that permits only minimal violation of the constraint:

$$(2) \text{Filter}_i(\text{Set}) = \{R \in \text{Set} : C_i(R) \text{ is minimal}\}$$

Given an underlying phonological input, its set of legal surface forms under the grammar—typically of size 1—is just

$$(3) \text{Filter}_n(\dots \text{Filter}_2(\text{Filter}_1(\mathbf{Gen}(\text{input}))))$$

where the function **Gen** is fixed across languages and $\mathbf{Gen}(\text{input}) \subseteq \mathbf{Repns}$ is a potentially infinite set of candidate surface forms.

In practice, each surface form in $\mathbf{Gen}(\text{input})$ must contain a silent copy of *input*, so the constraints can score it on how closely its pronounced material matches *input*. The constraints also score other criteria, such as how easy the material is to pronounce. If C_1 in a given language is violated by just the forms with coda consonants, then $\text{Filter}_1(\mathbf{Gen}(\text{input}))$ includes only coda-free candidates—regardless of their other demerits, such as discrepancies from *input* or unusual syllable structure. The remaining constraints are satisfied only as well as they can be given this set of survivors. Thus, when it is impossible to satisfy all constraints at once, successive filtering means early constraints take priority.

Questions under the new paradigm include these:

- *Generation*. How to implement the input-output mapping in (3)? A brute-force approach

fails to terminate if **Gen** produces infinitely many candidates. Speakers must solve this problem. So must linguists, if they are to know what their proposed grammars predict.

- *Comprehension.* How to invert the input-output mapping in (3)? Hearers must solve this.
- *Learning.* How to induce a lexicon and a phonology like (1) for a particular language, given the kind of evidence available to child language learners?

None of these questions is well-posed without restrictions on **Gen** and **Con**.

In the absence of such restrictions, computational linguists have assumed convenient ones. Ellison (1994) solves the generation problem where **Gen** produces a regular set of strings and **Con** admits all finite state transducers that can map a string to a number in unary notation. (Thus $C_i(R) = 4$ if the C_i transducer outputs the string 1111 on input R .) Tesar (1995, 1996) extends this result to the case where **Gen**(*input*) is the set of parse trees for *input* under some context-free grammar (CFG).¹ Tesar’s constraints are functions on parse trees such that $C_i([A [B_1 \dots] [B_2 \dots]])$ can be computed from $A, B_1, B_2, C_i(B_1)$, and $C_i(B_2)$. The optimal tree can then be found with a standard dynamic-programming chart parser for weighted CFGs.

It is an important question whether these formalisms are useful in practice. On the one hand, are they expressive enough to describe real languages? On the other, are they restrictive enough to admit good comprehension and unsupervised-learning algorithms?

The present paper sketches **primitive Optimality Theory** (OTP)—a new formalization of OT that is explicitly proposed as a linguistic hypothesis. Representations are autosegmental, **Gen** is trivial, and only certain simple and phonologically local constraints are allowed. I then show the following:

1. *Good news:* Generation in OTP can be solved attractively with finite-state methods. The solution is given in some detail.
2. *Good news:* OTP usefully restricts the space of grammars to be learned. (In particular, Generalized Alignment is outside the scope of finite-state or indeed context-free methods.)
3. *Bad news:* While OTP generation is close to linear on the size of the input form, it is NP-hard on the size of the grammar, which for human languages is likely to be quite large.
4. *Good news:* Ellison’s algorithm can be improved so that its exponential blowup is often avoided.

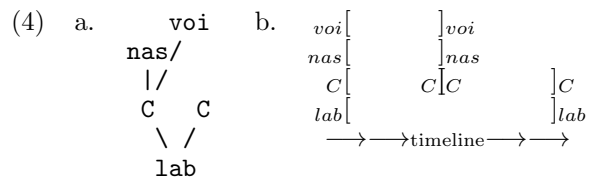
¹This extension is useful for OT syntax but may have little application to phonology, since the context-free case reduces to the regular case (i.e., Ellison) unless the CFG contains recursive productions.

2 Primitive Optimality Theory

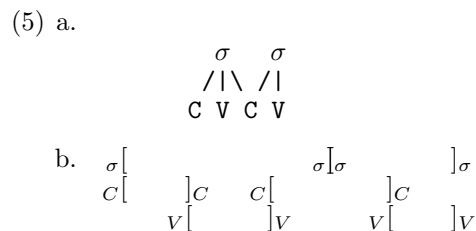
Primitive Optimality Theory, or OTP, is a formalization of OT featuring a homogeneous output representation, extremely local constraints, and a simple, unrestricted **Gen**. Linguistic arguments for OTP’s constraints and representations are given in (Eisner, 1997), whereas the present description focuses on its formal properties and suitability for computational work. An axiomatic treatment is omitted for reasons of space. Despite its simplicity, OTP appears capable of capturing virtually all analyses found in the (phonological) OT literature.

2.1 Repns: Representations in OTP

To represent [mp], OTP uses not the autosegmental representation in (4a) (Goldsmith, 1976; Goldsmith, 1990) but rather the simplified autosegmental representation in (4b), which has no association lines. Similarly (5a) is replaced by (5b). The central representational notion is that of a **constituent timeline**: an infinitely divisible line along on which constituents are laid out. Every constituent has width and edges.



For phonetic interpretation: $]_{\text{voi}}$ says to end voicing (laryngeal vibration). At the same instant, $]_{\text{nas}}$ says to end nasality (raise velum).



A timeline can carry the full panoply of phonological and morphological constituents—anything that phonological constraints might have to refer to. Thus, a timeline bears not only autosegmental features like nasal gestures [*nas*] and prosodic constituents such as syllables [σ], but also stress marks [\mathbf{x}], feature domains such as [*ATRdom*] (Cole & Kisseberth, 1994) and morphemes such as [*Stem*]. All these constituents are formally identical: each marks off an interval on the timeline. Let **Tiers** denote the fixed finite set of constituent types, $\{\text{nas}, \sigma, \mathbf{x}, \text{ATRdom}, \text{Stem}, \dots\}$.

It is always possible to recover the old representation (4a) from the new one (4b), under the convention that two constituents on the timeline are linked if their interiors overlap (Bird & Ellison, 1994). The **interior** of a constituent is the open interval that

excludes its edges. Thus, *lab* is linked to both consonants C in (4b), but the two consonants are not linked to each other, because their interiors do not overlap.

By eliminating explicit association lines, OTP eliminates the need for faithfulness constraints on them, or for well-formedness constraints against gapping or crossing of associations. In addition, OTP can refer naturally to the edges of syllables (or morphemes). Such edges are tricky to define in (5a), because a syllable’s features are scattered across multiple tiers and perhaps shared with adjacent syllables.

In diagrams of timelines, such as (4b) and (5b), the intent is that only horizontal order matters. Horizontal spacing and vertical order are irrelevant. Thus, a timeline may be represented as a finite collection S of labeled edge brackets, equipped with ordering relations \prec and \simeq that indicate which brackets precede each other or fall in the same place.

Valid timelines (those in **Repns**) also require that edge brackets come in matching pairs, that constituents have positive width, and that constituents of the same type do not overlap (i.e., two constituents on the same tier may not be linked).

2.2 Gen: Input and output in OTP

OT’s principle of Containment (Prince & Smolensky, 1993) says that each of the potential outputs in **Repns** includes a silent copy of the input, so that constraints evaluating it can consider the goodness of match between input and output. Accordingly, OTP represents both input and output constituents on the constituent timeline, but on different tiers. Thus surface nasal autosegments are bracketed with $nas[$ and $]nas$, while underlying nasal autosegments are bracketed with $\underline{nas}[$ and $]\underline{nas}$. The underlining is a notational convention to denote input material. No connection is required between $[nas]$ and $[\underline{nas}]$ except as enforced by constraints that prefer $[nas]$ and $[\underline{nas}]$ or their edges to overlap in some way. (6) shows a candidate in which underlying $[\underline{nas}]$ has surfaced “in place” but with rightward spreading.

$$(6) \quad \begin{array}{c} nas[\quad \quad \quad]nas \\ \underline{nas}[\quad \quad]\underline{nas} \end{array}$$

Here the left edges and interiors overlap, but the right edges fail to. Such overlap of interiors may be regarded as featural Input-Output Correspondence in the sense of (McCarthy & Prince, 1995).

The lexicon and morphology supply to **Gen** an **underspecified timeline**—a partially ordered collection of *input* edges. The use of a *partial* ordering allows the lexicon and morphology to supply floating tones, floating morphemes and templatic morphemes.

Given such an underspecified timeline as lexical input, **Gen** outputs the set of *all* fully specified timelines that are consistent with it. No new input constituents may be added. In essence, **Gen** generates

every way of refining the partial order of input constituents into a total order and decorating it freely with output constituents. Conditions such as the prosodic hierarchy (Selkirk, 1980) are enforced by universally high-ranked constraints, not by **Gen**.²

2.3 Con: The primitive constraints

Having described the representations used, it is now possible to describe the constraints that evaluate them. OTP claims that **Con** is restricted to the following two families of **primitive constraints**:

- (7) $\alpha \rightarrow \beta$ (“implication”):
“Each α temporally overlaps some β .”

Scoring: $\text{Constraint}(R) =$ number of α ’s in R that do not overlap any β .

- (8) $\alpha \perp \beta$ (“clash”):
“Each α temporally overlaps no β .”

Scoring: $\text{Constraint}(R) =$ number of (α, β) pairs in R such that the α overlaps the β .

That is, $\alpha \rightarrow \beta$ says that α ’s attract β ’s, while $\alpha \perp \beta$ says that α ’s repel β ’s. These are simple and arguably natural constraints; no others are used.

In each primitive constraint, α and β each specify a phonological event. An event is defined to be either a type of labeled edge, written e.g. $\sigma[$, or the interior (excluding edges) of a type of labeled constituent, written e.g. σ . To express some constraints that appear in real phonologies, it is also necessary to allow α and β to be non-empty conjunctions and disjunctions of events. However, it appears possible to limit these cases to the forms in (9)–(10). Note that other forms, such as those in (11), can be decomposed into a sequence of two or

²The formalism is complicated slightly by the possibility of deleting segments (syncope) or inserting segments (epenthesis), as illustrated by the candidates below.

- (i) Syncope ($CVC \Rightarrow CC$): the \underline{V} is crushed to zero width so the \underline{C} ’s can be adjacent.

$$\begin{array}{c} c[\quad \quad]c \quad \quad]c \\ \underline{c}[\quad \quad]\underline{c} \quad \quad]\underline{c} \\ \underline{v}|\underline{v} \end{array}$$

- (ii) Epenthesis ($CC \Rightarrow CVC$): the \underline{C} ’s are pushed apart.

$$\begin{array}{c} \quad \quad \quad v[\quad \quad]v \\ c[\quad \quad]c \quad \quad c[\quad \quad]c \\ \underline{c}[\quad \quad]\underline{c} \quad \quad \underline{c}[\quad \quad]\underline{c} \end{array}$$

In order to allow adjacency of the surface consonants in (i), as expected by assimilation processes (and encouraged by a high-ranked constraint), note that the underlying vowel must be allowed to have zero width—an option available to input but not output constituents. The input representation must specify only $\underline{v}[\preceq]\underline{v}$, not $\underline{v}[\prec]\underline{v}$. Similarly, to allow (ii), the input representation must specify only $]\underline{c}_1 \preceq \underline{c}_2[$, not $]\underline{c}_1 \simeq \underline{c}_2[$.

more constraints.³

- (9) $(\alpha_1 \text{ and } \alpha_2 \text{ and } \dots) \rightarrow (\beta_1 \text{ or } \beta_2 \text{ or } \dots)$
Scoring: $\text{Constraint}(R) =$ number of vectors of events $\langle A_1, A_2, \dots \rangle$ of types $\alpha_1, \alpha_2, \dots$ respectively that all overlap on the timeline and whose intersection does not overlap any event of type β_1 or β_2 or \dots .
- (10) $(\alpha_1 \text{ and } \alpha_2 \text{ and } \dots) \perp (\beta_1 \text{ and } \beta_2 \text{ and } \dots)$
Scoring: $\text{Constraint}(R) =$ number of vectors of events $\langle A_1, A_2, \dots, B_1, B_2, \dots \rangle$ of types $\alpha_1, \alpha_2, \dots, \beta_1, \beta_2, \dots$ respectively that all overlap on the timeline.
 (Could also be notated:
 $\alpha_1 \perp \alpha_2 \perp \dots \perp \beta_1 \perp \beta_2 \perp \dots$)
- (11) $\alpha \rightarrow (\beta_1 \text{ and } \beta_2)$ [cf. $\alpha \rightarrow \beta_1 \gg \alpha \rightarrow \beta_2$]
 $(\alpha_1 \text{ or } \alpha_2) \rightarrow \beta$ [cf. $\alpha_1 \rightarrow \beta \gg \alpha_2 \rightarrow \beta$]

The unifying theme is that each primitive constraint counts the number of times a candidate gets into some bad *local* configuration. This is an interval on the timeline throughout which certain events (one or more specified edges or interiors) are all present and certain other events (zero or more specified edges or interiors) are all absent.

Several examples of phonologically plausible constraints, with monikers and descriptions, are given below. (Eisner, 1997) shows how to rewrite hundreds of constraints from the literature in the primitive constraint notation, and discusses the problematic case of reduplication. (Eisner, in press) gives a detailed stress typology using only primitive constraints; in particular, non-local constraints such as FTBIN, FOOTFORM, and Generalized Alignment (McCarthy & Prince, 1993) are eliminated.

- (12) a. ONSET: $\sigma[\rightarrow C[$
 “Every syllable starts with a consonant.”
- b. NONFINALITY: $]_{Word} \perp]_F$
 “The end of a word may not be footed.”
- c. FTSYL: $F[\rightarrow \sigma[\ ; \]_F \rightarrow]_\sigma$
 “Feet start and end on syllable boundaries.”
- d. PACKFEET: $]_F \rightarrow F[$
 “Each foot is followed immediately by another foot; i.e., minimize the number of gaps between feet. Note that the final foot, if any, will always violate this constraint.”
- e. NOCLASH: $]_{\mathbf{x}} \perp]_{\mathbf{x}}[$
 “Two stress marks may not be adjacent.”
- f. PROGRESSIVEVOICING: $]_{voi} \perp C[$
 “If the segment preceding a consonant is voiced, voicing may not stop prior to the

³Such a sequence does alter the meaning slightly. To get the exact original meaning, we would have to decompose into so-called “unranked” constraints, whereby $C_i(R)$ is defined as $C_{i_1}(R) + C_{i_2}(R)$. But such ties undermine OT’s idea of strict ranking: they confer the power to minimize linear functions such as $(C_1 + C_1 + C_1 + C_2 + C_3 + C_3)(R) = 3C_1(R) + C_2(R) + 2C_3(R)$. For this reason, OTP currently disallows unranked constraints; I know of no linguistic data that crucially require them.

- consonant but must be spread onto it.”
- g. NASVOI: $nas \rightarrow voi$
 “Every nasal gesture must be at least partly voiced.”
- h. FULLNASVOI: $nas \perp voi[\ ; \]_{voi}$
 “A nasal gesture may not be only partly voiced.”
- i. MAX(voi) or PARSE(voi): $voi \rightarrow voi$
 “Underlying voicing features surface.”
- j. DEP(voi) or FILL(voi): $voi \rightarrow voi$
 “Voicing features appear on the surface only if they are also underlying.”
- k. NOSPREADRIGHT(voi): $voi \perp]_{voi}$
 “Underlying voicing may not spread rightward as in (6).”
- l. NONDEGENERATE: $F \rightarrow \mu[$
 “Every foot must cross at least one mora boundary $\mu[$.”
- m. TAUTOMORPHICFOOT: $F \perp]_{Morph}$
 “No foot may cross a morpheme boundary.”

3 Finite-state generation in OTP

3.1 A simple generation algorithm

Recall that the generation problem is to find the output set S_n , where

- (13) a. $S_0 = \mathbf{Gen}(\text{input}) \subseteq \mathbf{Repns}$
 b. $S_{i+1} = \text{Filter}_{i+1}(S_i) \subseteq S_i$

Since in OTP, the input is a partial order of edge brackets, and S_n is a set of one or more total orders (timelines), a natural approach is to successively refine a partial order. This has merit. However, not every S_i can be represented as a single partial order, so the approach is quickly complicated by the need to encode disjunction.

A simpler approach is to represent S_i (as well as *input* and **Repns**) as a finite-state automaton (FSA), denoting a regular set of strings that encode timelines. The idea is essentially due to (Ellison, 1994), and can be boiled down to two lines:

- (14) **Ellison’s algorithm** (variant).
 $S_0 = \text{input} \cap \mathbf{Repns}$
 = all conceivable outputs containing *input*
 $S_{i+1} = \text{BestPaths}(S_i \cap C_{i+1})$

Each constraint C_i must be formulated as an *edge-weighted FSA* that scores candidates: C_i accepts any string R , on a single path of weight $C_i(R)$.⁴ BestPaths is Dijkstra’s “single-source shortest paths” algorithm, a dynamic-programming algorithm that prunes away all but the minimum-weight paths in an automaton, leaving an unweighted automaton.

OTP is simple enough that it can be described in this way.⁵ The next section gives a nice encoding.

⁴Weighted versions of the state-labeled finite automata of (Bird & Ellison, 1994) could be used instead.

⁵The converse is also true, as was shown at the talk accompanying this paper: given enough abstract tiers, OTP can simulate any finite-state OT grammar.

Suppose the input is simply [*Stem*]. Filtering $\mathbf{Gen}(\text{input})$ through constraints (23a–e), we are left with just those candidates where *Stem* bears n disjoint constituents of type S , each coextensive with a constituent bearing a different label $v \in V(G)$. (These candidates satisfy (23a–d) but violate (23e) n times.) (23f) says that a chain of abutting constituents $[u[v[w] \dots]$ is allowed only if it corresponds to a path in G . Finally, (23g) forces the grammar to minimize the number of such chains. If the minimum is 1 (i.e., an arbitrarily selected output candidate violates (23g) only once), then G has a Hamilton path.

When confronted with this pathological case, the finite-state methods respond essentially by enumerating all possible permutations of $V(G)$ (though with sharing of prefixes). The machine state stores, among other things, the subset of $V(G)$ that has already been seen; so there are at least $2^{|\mathbf{Tiers}|}$ states.

It must be emphasized that if the grammar is *fixed in advance*, algorithm (14) is close to linear in the size of the input form: it is dominated by a constant number of calls to Dijkstra’s BestPaths method, each taking time $O(|\text{input arcs}| \log |\text{input states}|)$. There are nonetheless three reasons why the above result is important. (a) It raises the practical specter of huge constant factors ($> 2^{40}$) for real grammars. Even if a fixed grammar can somehow be compiled into a fast form for use with many inputs, the compilation itself will have to deal with this constant factor. (b) The result has the interesting implication that candidate sets can arise that cannot be concisely represented with FSAs. For if all S_i were polynomial-sized in (14), the algorithm would run in polynomial time. (c) Finally, the grammar is *not* fixed in all circumstances: both linguists and children crucially experiment with different theories.

4.3 Work in progress: Factored automata

The previous section gave a useful trick for speeding up Ellison’s algorithm in the typical case. We are currently experimenting with additional improvements along the same lines, which attempt to defer intersection by keeping tiers separate as long as possible.


The idea is to represent the candidate set S_i not as a large unweighted FSA, but rather as a collection \mathcal{A} of preferably small unweighted FSAs, called **factors**, each of which mentions as few tiers as possible. This collection, called a **factored automaton**, serves as a compact representation of $\cap \mathcal{A}$. It usually has far fewer states than $\cap \mathcal{A}$ would if the intersection were carried out.

For instance, the natural factors of S_0 are *input* and all the tier rules (see 18). This requires only $O(|\mathbf{Tiers}| + |\text{input}|)$ states, not $O(2^{|\mathbf{Tiers}|} \cdot |\text{input}|)$. Using factored automata helps Ellison’s algorithm (14) in several ways:

- The candidate sets S_i tend to be represented

more compactly.

- In (14), the constraint C_{i+1} needs to be intersected with only certain factors of S_i .
- Sometimes C_{i+1} does not need to be intersected with the input, because they do not mention any of the same tiers. Then step $i + 1$ can be performed in time independent of input length.

Example: *input* = , which is a 43-state automaton, and C_1 is $F \rightarrow \mathbf{x}$, which says that every foot bears a stress mark. Then to find $S_1 = \text{BestPaths}(S_0 \cap C_1)$, we need only consider S_0 ’s tier rules for F and \mathbf{x} , which require well-formed feet and well-formed stress marks, and combine them with C_1 to get a new factor that requires stressed feet. No other factors need be involved.

The key operation in (14) is to find $\text{Bestpaths}(\mathcal{A} \cap C)$, where \mathcal{A} is an unweighted factored automaton representing a candidate set, and C is an ordinary weighted FSA (a constraint). This is the **best intersection** problem. For concreteness let us suppose that C encodes $F \rightarrow \mathbf{x}$, a two-state constraint.

A naive idea is simply to add $F \rightarrow \mathbf{x}$ to \mathcal{A} as a new factor. However, this ignores the BestPaths step: we wish to keep just the best paths in $F \rightarrow \mathbf{x}$ that are compatible with \mathcal{A} . Such paths might be long and include unrolled cycles in $F \rightarrow \mathbf{x}$. For example, a weight-1 path would describe a chain of optimal stressed feet interrupted by a single unstressed one where \mathcal{A} happens to block stress.

A corrected variant is to put $I = \cap \mathcal{A}$ and run BestPaths on $I \cap C$. Let the pruned result be B . We could add B directly back to \mathcal{A} as a new factor, but it is large. We would rather add a smaller factor B' that has the same effect, in that $I \cap B' = I \cap B$. (B' will look something like the original C , but with some paths missing, some states split, and some cycles unrolled.) Observe that each state of B has the form $i \times c$ for some $i \in I$ and $c \in C$. We form B' from B by “re-merging” states $i \times c$ and $i' \times c$ where possible, using an approach similar to DFA minimization.

Of course, this variant is not very efficient, because it requires us to find and use $I = \cap \mathcal{A}$. What we really want is to follow the above idea but use a smaller I , one that considers just the relevant factors in \mathcal{A} . We need not consider factors that will not affect the choice of paths in C above.

Various approaches are possible for choosing such an I . The following technique is completely general, though it may or may not be practical.

Observe that for BestPaths to do the correct thing, I needs to reflect the sum total of \mathcal{A} ’s constraints on F and \mathbf{x} , the tiers that C mentions. More formally, we want I to be the projection of the candidate set $\cap \mathcal{A}$ onto just the F and \mathbf{x} tiers. Unfortunately, these constraints are not just reflected in the factors mentioning F or \mathbf{x} , since the allowed configurations of F and \mathbf{x} may be mediated through

additional factors. As an example, there may be a factor mentioning F and ψ , some of whose paths are incompatible with the input factor, because the latter allows ψ only in certain places or because only allows paths of length 14.

1. Number the tiers such that F and \mathbf{x} are numbered 0, and all other tiers have distinct positive numbers.
2. Partition the factors of \mathcal{A} into lists $L_0, L_1, L_2, \dots, L_k$, according to the highest-numbered tier they mention. (Any factor that mentions no tiers at all goes onto L_0 .)
3. If $k = 0$, then return $\cap L_k$ as our desired I .
4. Otherwise, $\cap L_k$ exhausts tier k 's ability to mediate relations among the factors. Modify the arc labels of $\cap L_k$ so that they no longer restrict (mention) k . Then add a determinized, minimized version of the result to L_j , where j is the highest-numbered tier it now mentions.
5. Decrement k and return to step 3.

If $\cap \mathcal{A}$ has k factors, this technique must perform $k - 1$ intersections, just as if we had put $I = \cap \mathcal{A}$. However, it intersperses the intersections with determinization and minimization operations, so that the automata being intersected tend not to be large. In the best case, we will have $k - 1$ intersection-determinization-minimizations that cost $O(1)$ apiece, rather than $k - 1$ intersections that cost up to $O(2^k)$ apiece.

5 Conclusions

Primitive Optimality Theory, or OTP, is an attempt to produce a simple, rigorous, constraint-based model of phonology that is closely fitted to the needs of working linguists. I believe it is worth study both as a hypothesis about Universal Grammar and as a formal object.

The present paper introduces the OTP formalization to the computational linguistics community. We have seen two formal results of interest, both having to do with generation of surface forms:

- OTP's *generative power* is low: finite-state optimization. In particular it is more constrained than theories using Generalized Alignment. This is good news for comprehension and learning.
- OTP's *computational complexity*, for generation, is nonetheless high: NP-hard on the size of the grammar. This is mildly unfortunate for OTP and for the OT approach in general.

It remains true that for a fixed grammar, the time to do generation is close to linear on the size of the input (Ellison, 1994), which is heartening if we intend to optimize long utterances with respect to a fixed phonology.

Finally, we have considered the prospect of building a practical tool to generate optimal outputs from OT theories. We saw above to set up the representations and constraints efficiently using deterministic finite-state automata, and how to remedy some hidden inefficiencies in the seminal work of (Ellison, 1994), achieving at least a 100-fold observed speedup. Delayed intersection and aggressive pruning prove to be important. Aggressive minimization and a more compact, "factored" representation of automata may also turn out to help.

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