

Distributional Learning of Context-Free Grammars.

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UCL

Outline

Introduction

Weak Learning

Strong Learning

An Algebraic Theory of CFGs

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Strong Learning

An Algebraic Theory of CFGs

Machine learning

Standard machine learning problem

We learn a function $f : \mathcal{X} \rightarrow \mathcal{Y}$ from a sequence of input-output pairs $\langle (\mathbf{x}_1, \mathbf{y}_1) \dots (\mathbf{x}_n, \mathbf{y}_n) \rangle$

Convergence

As $n \rightarrow \infty$ we want our hypothesis \hat{f} to tend to f
Ideally we want $\hat{f} = f$.

Vector spaces

Standard two assumptions

1. Assume sets have some algebraic structure:
 - ▶ \mathcal{X} is \mathbb{R}^n
 - ▶ \mathcal{Y} is \mathbb{R}
2. Assume f satisfies some smoothness assumptions:
 - ▶ f is linear
 - ▶ or satisfies some Lipschitz condition: $|f(\mathbf{x}_i) - f(\mathbf{x}_j)| \leq c|\mathbf{x}_i - \mathbf{x}_j|$

- ▶ The input examples are strings.
- ▶ No output (unsupervised learning!)
- ▶ Our representations are context-free grammars.

Context-Free Grammars

Context-Free Grammar

$$G = \langle \Sigma, V, S, P \rangle$$

$$\mathcal{L}(G, A) = \{w \in \Sigma^* \mid A \xRightarrow{*}_G w\}$$

Example

$$\Sigma = \{a, b\}, V = \{S\}$$

$$P = \{S \rightarrow ab, S \rightarrow aSb, S \rightarrow \epsilon\}$$

$$\mathcal{L}(G, S) = \{a^n b^n \mid n \geq 0\}$$

Least fixed point semantics

[Ginsburg and Rice(1962)]

Interpret this as a set of equations in $\mathcal{P}(\Sigma^*)$

$$S = (a \circ b) \vee (a \circ S \circ b) \vee \epsilon$$

Least fixed point semantics

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Interpret this as a set of equations in $\mathcal{P}(\Sigma^*)$

$$S = (a \circ b) \vee (a \circ S \circ b) \vee \epsilon$$

- ▶ Ξ is the set of functions $V \rightarrow \mathcal{P}(\Sigma^*)$
- ▶ $\Phi_G : \Xi \rightarrow \Xi$

$$\Phi_G(\xi)[S] = (a \circ b) \vee (a \circ \xi(S) \circ b) \vee \epsilon$$

Least fixed point $\xi_G = \bigvee_n \Phi_G^n(\xi_\perp) = \{S \rightarrow \mathcal{L}(G, S)\}$

What Algebra?

Monoid: $\langle S, \circ, 1 \rangle$

Σ^*

What Algebra?

Monoid: $\langle S, \circ, 1 \rangle$

Σ^*

Complete Idempotent Semiring: $\langle S, \circ, 1, \vee, \perp \rangle$

$\mathcal{P}(\Sigma^*)$

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Running example

Propositional logic

Alphabet

| | |
|-------------------------------|--|
| rain, snow, hot, cold, danger | A_1, A_2, \dots |
| and, or, implies, iff | $\wedge, \vee, \rightarrow, \leftrightarrow$ |
| not | \neg |
| open, close | $(,)$ |

Running example

Propositional logic

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| open, close | $(,)$ |

- ▶ rain
- ▶ open snow implies cold close
- ▶ open snow implies open not hot close close

Distributional Learning

[Harris(1964)]

- ▶ Look at the dog
- ▶ Look at the cat

Distributional Learning

[Harris(1964)]

- ▶ Look at the dog
- ▶ Look at the cat
- ▶ That cat is crazy

Distributional Learning

[Harris(1964)]

- ▶ Look at the dog
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English counterexample

- ▶ I can swim
- ▶ I may swim
- ▶ I want a can of beer

English counterexample

- ▶ I can swim
- ▶ I may swim
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- ▶ *I want a may of beer

English counterexample

- ▶ She is Italian
- ▶ She is a philosopher
- ▶ She is an Italian philosopher

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- ▶ She is Italian
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Logic example

Propositional logic is *substitutable*:

- ▶ open rain **and** cold close
- ▶ open rain **implies** cold close
- ▶ open snow **implies** open not hot close

Logic example

Propositional logic is *substitutable*:

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- ▶ open snow **implies** open not hot close
- ▶ open snow **and** open not hot close

Formally

The Syntactic Congruence: a monoid congruence

Two nonempty strings u, v are congruent ($u \equiv_L v$) if for all $l, r \in \Sigma^*$

$$lur \in L \Leftrightarrow lvr \in L$$

We write $[u]$ for the congruence class of u .

Definition

L is substitutable if

$$lur \in L, lvr \in L \Rightarrow u \equiv_L v$$

Example

Input data $D \subseteq L$

- ▶ hot
- ▶ cold
- ▶ open hot or cold close
- ▶ open not hot close
- ▶ open hot and cold close
- ▶ open hot implies cold close
- ▶ open hot iff cold close
- ▶ danger
- ▶ rain
- ▶ snow

One production for each example

- ▶ $S \rightarrow \text{hot}$
- ▶ $S \rightarrow \text{cold}$
- ▶ $S \rightarrow \text{open hot or cold close}$
- ▶ $S \rightarrow \text{open not hot close}$
- ▶ $S \rightarrow \text{open hot and cold close}$
- ▶ $S \rightarrow \text{open hot implies cold close}$
- ▶ $S \rightarrow \text{open hot iff cold close}$
- ▶ $S \rightarrow \text{danger}$
- ▶ $S \rightarrow \text{rain}$
- ▶ $S \rightarrow \text{snow}$

A trivial grammar

Input data D

$D = \{w_1, w_2, \dots, w_n\}$ are nonempty strings.

Starting grammar

$S \rightarrow w_1, S \rightarrow w_2, \dots, S \rightarrow w_n$

$\mathcal{L}(G) = D$

A trivial grammar

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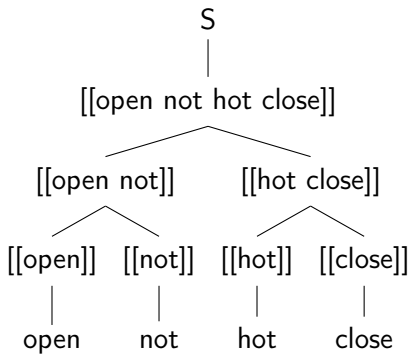
$\mathcal{L}(G) = D$

Binarise this every way

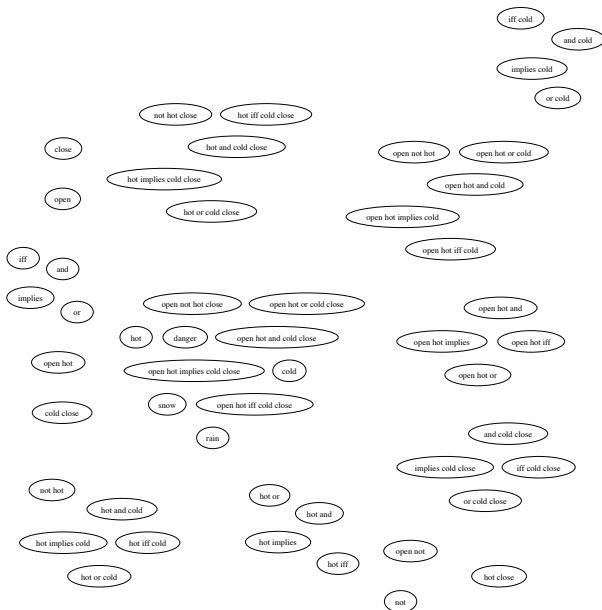
One nonterminal $[[w]]$ for every substring w .

- ▶ $[[a]] \rightarrow a$
- ▶ $S \rightarrow \{w\}, w \in D$
- ▶ $[[w]] \rightarrow [[u]][[v]]$ when $w = u \cdot v$

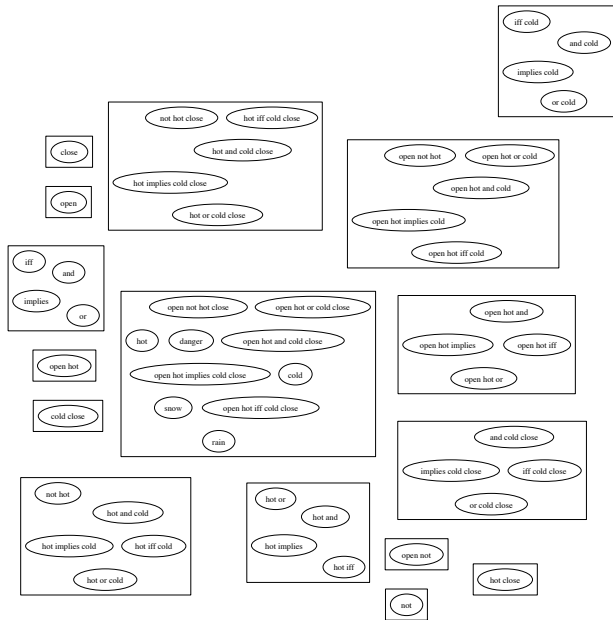
$$\mathcal{L}(G, [[w]]) = \{w\}$$



Nonterminal for each substring



Nonterminal for each cluster



Productions

Observation

If $w = u \cdot v$ then $[w] \supseteq [u] \cdot [v]$

Productions

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Add production

$[[w]] \rightarrow [[u]][[v]]$

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Add production

$[[w]] \rightarrow [[u]][[v]]$

Consequence

If L is substitutable, then

$$\mathcal{L}(G, [[w]]) \subseteq [w]$$

$$\mathcal{L}(G) \subseteq L$$

Theorem [Clark and Eyraud(2007)]

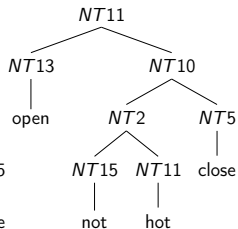
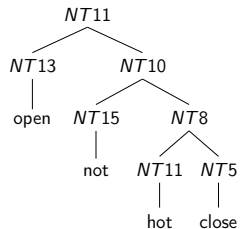
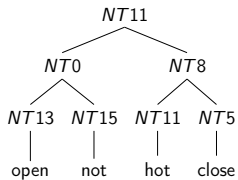
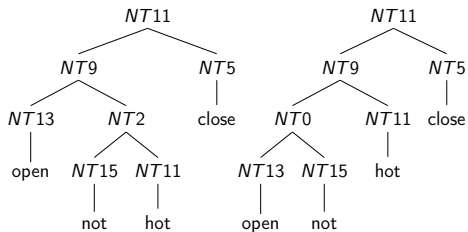
- ▶ If the language is a substitutable context-free language, then the hypothesis grammar will converge to a correct grammar.
- ▶ Efficient; provably correct

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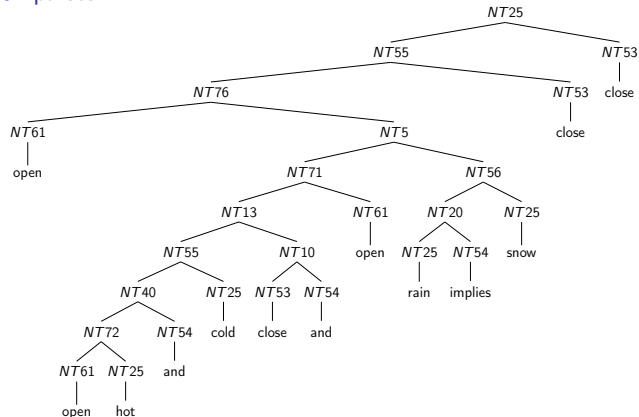
But the grammar may be different for each input data set!

open not hot close



open open hot and cold close and open rain implies snow
close close

327204 parses



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Strong Learning

An Algebraic Theory of CFGs

Strong Learning

Target class of grammars

\mathcal{G} is some set of context-free grammars.

Pick some grammar $G_* \in \mathcal{G}$

Weak learning

We receive examples w_1, \dots, w_n, \dots

We produce a series of hypotheses G_1, \dots, G_n, \dots

We want G_n to converge to some grammar \hat{G} such that

$$L(\hat{G}) = L(G_*)$$

Strong Learning

Target class of grammars

\mathcal{G} is some set of context-free grammars.

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Strong learning

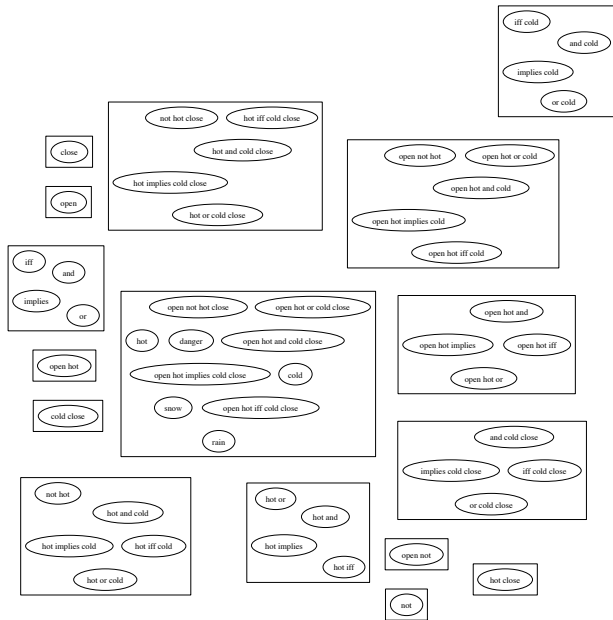
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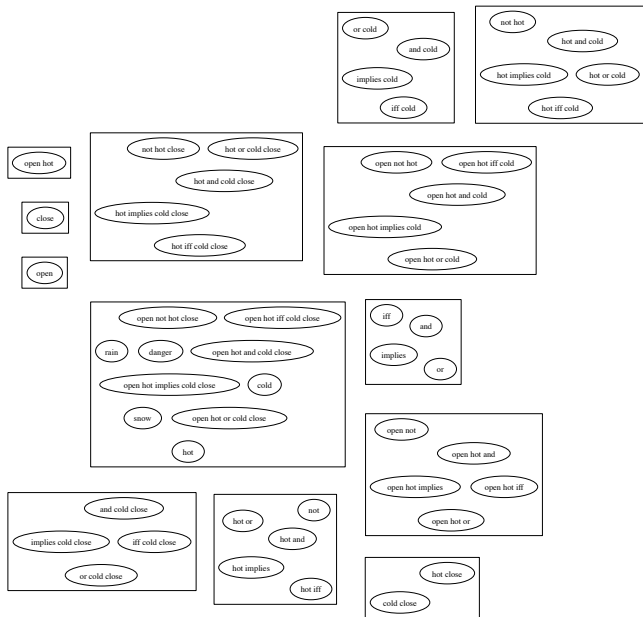
We want G_n to converge to some grammar \hat{G} such that

$\hat{G} \equiv G_*$

Inaccurate clusters



Correct congruence classes



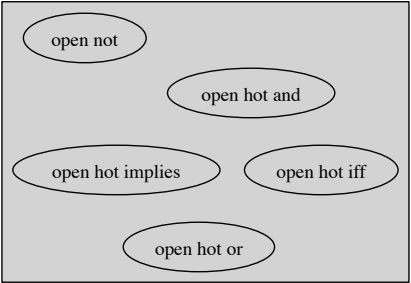
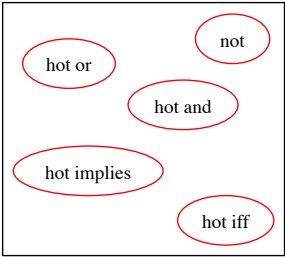
Myhill-Nerode Theorem

A language has a finite number of congruence classes if and only if it is regular.

Myhill-Nerode Theorem

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We need some principled way of picking a finite collection of “good” congruence classes.



Definition

Definition

A congruence class X is composite if there are two congruence classes Y, Z such that $X = YZ$.

(and neither Y nor Z is the class $[\lambda]$)

Definition

A congruence class X is prime if it is not composite.

The whole is greater than the sum of the parts

open hot

close

open

not hot close hot or cold close
hot and cold close
hot implies cold close
hot iff cold close

or cold
and cold
implies cold
iff cold

not hot
hot and cold
hot implies cold hot or cold
hot iff cold

open not hot open hot iff cold
open hot and cold
open hot implies cold
open hot or cold

open not hot close open hot iff cold close
rain danger open hot and cold close
open hot implies cold close cold
snow open hot or cold close
hot

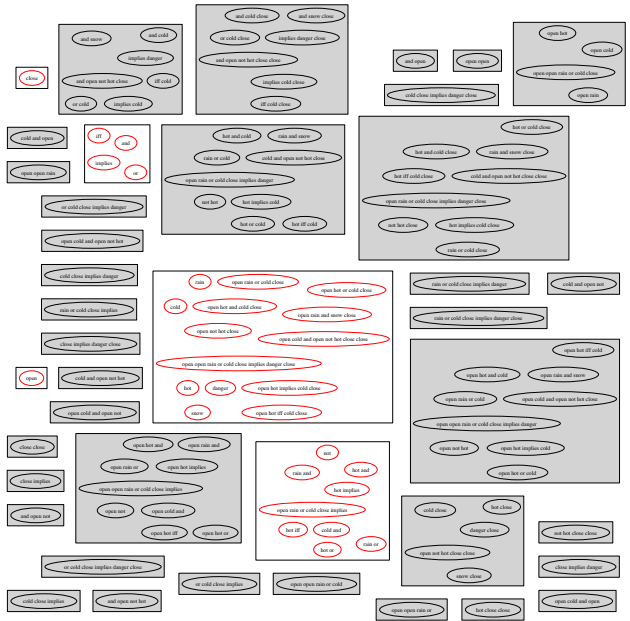
iff and
implies or

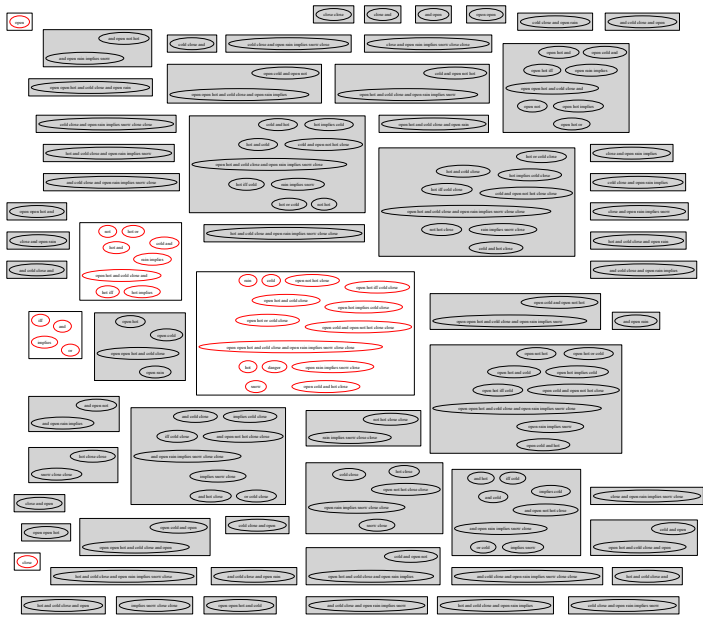
open not
open hot and
open hot implies open hot iff
open hot or

and cold close
implies cold close iff cold close
or cold close

hot or not
hot and
hot implies
hot iff

hot close
cold close





Restriction

- ▶ We only consider substitutable languages which have a finite number of primes.
- ▶ We define nonterminals only for these primes.

| Label | Examples |
|-------|--------------------------------------|
| P | rain, cold, open rain and cold close |
| O | open |
| C | close |
| B | and, or, ... |
| N | not, hot or, cold and ... |

Fundamental theorem of substitutable languages

Every congruence class Q can be uniquely represented as a sequence of primes such that $Q = P_1 \dots P_n$

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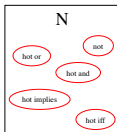
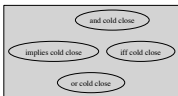
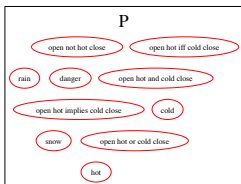
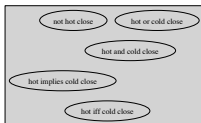
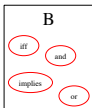
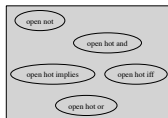
Intuition

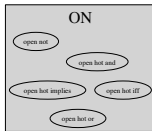
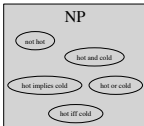
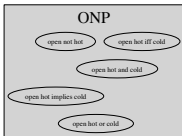
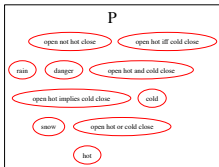
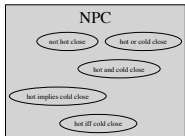
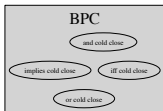
If $X = YZ$, and we have a rule

$$P \rightarrow QXR$$

then we can change it to

$$P \rightarrow QYZR$$





Productions

We need non-binary rules.

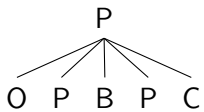
Correct productions

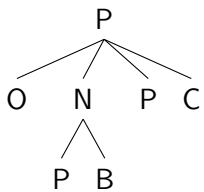
$$P_0 \rightarrow P_1 \dots P_k$$

where $P_0 \supsetneq P_1 \dots P_k$

Infinite number of correct productions

- ▶ $N \rightarrow PB$
- ▶ $P \rightarrow ONPC$
- ▶ $P \rightarrow OPBPC$
- ▶ $P \rightarrow ONONPCC$
- ▶ ...





Productions

Valid productions

- ▶ Correct productions where the right hand side does not contain the right hand side of a valid production.
- ▶ If there are n primes then there are at most n^2 valid productions.

Examples

- ▶ $N \rightarrow PB$
- ▶ $P \rightarrow ONPC$

A Strong Learning Result

Class of grammars

\mathcal{G}_{sc} is the class of canonical grammars for all substitutable languages with a finite number of primes.

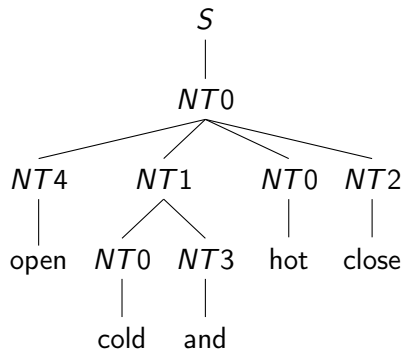
Theorem [Clark(2014)]

There is an algorithm which learns \mathcal{G}_{sc}

- ▶ From positive examples
- ▶ Identification in the limit
- ▶ Strongly: converges structurally
- ▶ Using polynomial time and data

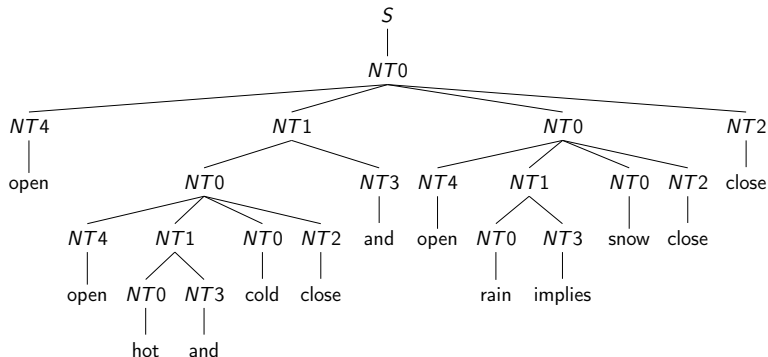
Running example

(verbatim output from implementation)



open open hot and cold close and open rain implies snow
close close

1 parse



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Contexts

A context is a string with a hole:

$$l \square r$$

Derivation contexts

The derivation contexts of CFGs are just string contexts:

$$S \xRightarrow{*}_G lNr$$

Definition

Filling the hole

$$l \square r \odot u = lur$$

A factorisation of a language L

C is a set of contexts; S is a set of strings

$$C \odot S \subseteq L$$

Context free grammars

Contexts and yields

$$\mathcal{L}(G, N) = \{w \in \Sigma^* \mid N \xRightarrow{*} w\}$$

$$\mathcal{C}(G, N) = \{I \square r \mid S \xRightarrow{*} INr\}.$$

Nonterminals in a context-free grammar

$$\mathcal{C}(G, N) \odot \mathcal{L}(G, N) \subseteq L$$

Context free grammars

Contexts and yields

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Nonterminals in a context-free grammar

$$\mathcal{C}(G, N) \odot \mathcal{L}(G, N) \subseteq L$$

We can reverse this process and go from a collection of decompositions back to a CFG.

Polar maps

If S is a set of strings:

$$S^\triangleright = \{l \square r \mid \forall u \in S, lur \in L\} \quad (1)$$

If C is a set of contexts:

$$C^\triangleleft = \{u \in \Sigma^* \mid \forall l \square r \in C, lur \in L\} \quad (2)$$

Polar maps

If S is a set of strings:

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If C is a set of contexts:

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\cdot^\triangleright and \cdot^\triangleleft form a Galois connection between sets of strings and sets of contexts.

Closed sets of strings

- ▶ $\cdot \triangleright \triangleleft$ is a closure operator on the sets of strings;
- ▶ $X^\triangleright = Y^\triangleright$ is a CIS-congruence;
- ▶ L is always closed.

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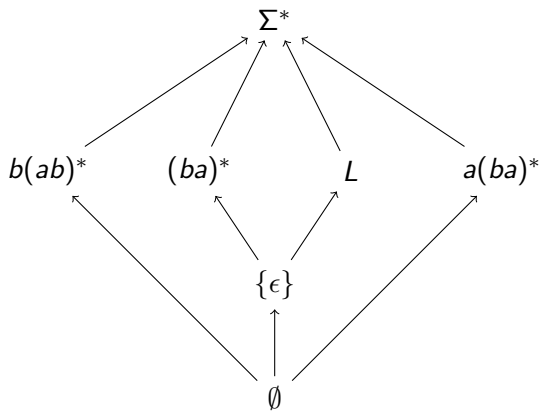
The syntactic concept lattice

The set of all closed sets of strings form a complete idempotent semiring: $\mathfrak{B}(L)$.

(A generalisation of the syntactic monoid; the collection of maximal decompositions into strings and contexts.)

L is regular iff $\mathfrak{B}(L)$ is finite

$$L = (ab)^*$$



Recognising a language

Definition

We say that a CIS B recognizes L if there is a surjective morphism h from $\mathcal{P}(\Sigma^*) \rightarrow B$ such that $h^*(h(L)) = L$, where h^* is the residual of h .

Recognising a language

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We say that a CIS B recognizes L if there is a surjective morphism h from $\mathcal{P}(\Sigma^*) \rightarrow B$ such that $h^*(h(L)) = L$, where h^* is the residual of h .

Given a CIS B and a homomorphism $h : \mathcal{P}(\Sigma^*) \rightarrow B$, we can define a new grammar $\phi_h(G)$ by merging nonterminals M, N if

$$h(\mathcal{L}(G, M)) = h(\mathcal{L}(G, N))$$

Theorem

Let G be a CFG over Σ and h a homomorphism $h : \mathcal{P}(\Sigma^*) \rightarrow B$.

Then

- ▶ $\phi_h(G)$ defines the same language as G iff
- ▶ B recognizes L through h

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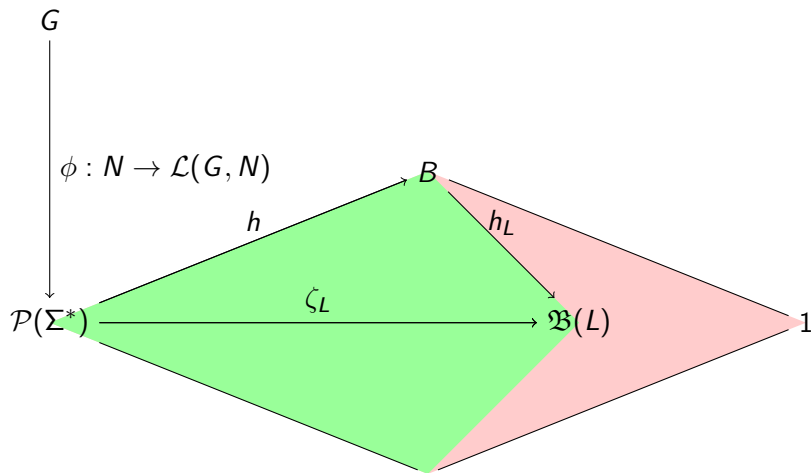
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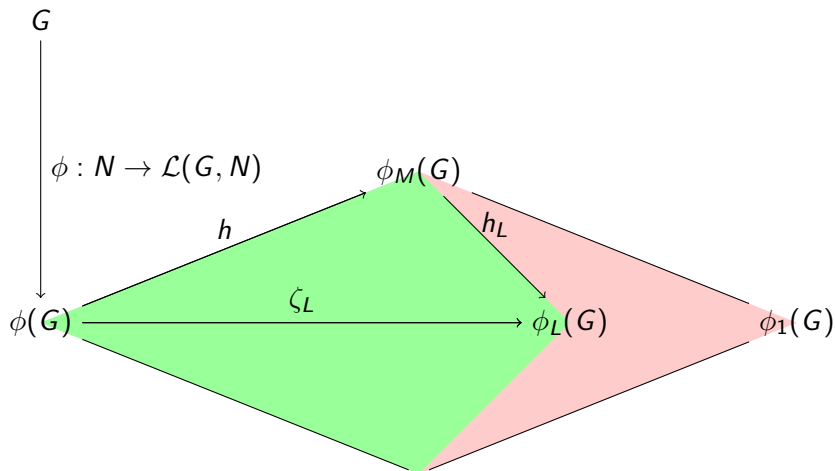
Uniqueness

There is a unique 'smallest' CIS that recognizes L : which is $\mathfrak{B}(L)$.

The universal morphism



The universal cfg-morphism



[Clark(2013)]

Mergeable nonterminals

If

$$\mathcal{L}(G, M)^{\triangleright\triangleleft} = \mathcal{L}(G, N)^{\triangleright\triangleleft}$$

then we can merge M and N without increasing the language defined by G ,

[Clark(2013)]

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Minimal grammars correspond to maximal factorisations

A grammar without mergeable nonterminals will have nonterminals that correspond to closed sets of strings.

Conclusion

- ▶ We can learn context-free grammars weakly decomposing strings into contexts and substrings.
- ▶ Minimal grammars will correspond to maximal decompositions.
- ▶ We can learn grammars strongly by identifying structure in some canonical algebras associated with the languages:
 - ▶ the syntactic monoid
 - ▶ the syntactic concept lattice.
- ▶ The same approach applies to Multiple Context-Free Grammars, a mildly context-sensitive grammar formalism.

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