

# Logical and Computational Structures for Linguistic Modeling

## Part 3 – Mildly Context-Sensitive Formalisms

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30 Septembre 2014

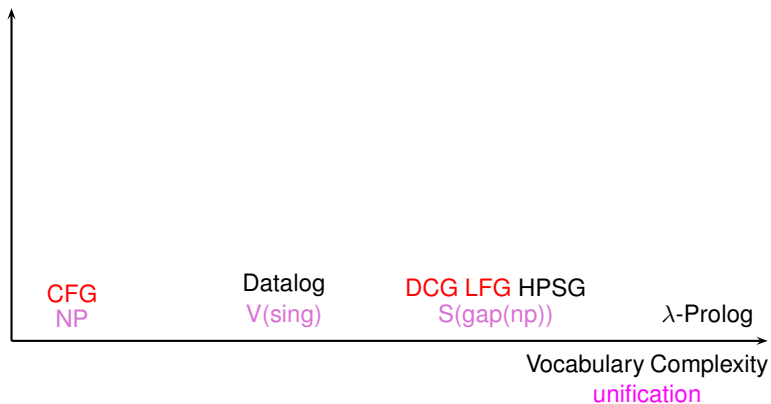
# Part I

## Tree Adjoining Grammars

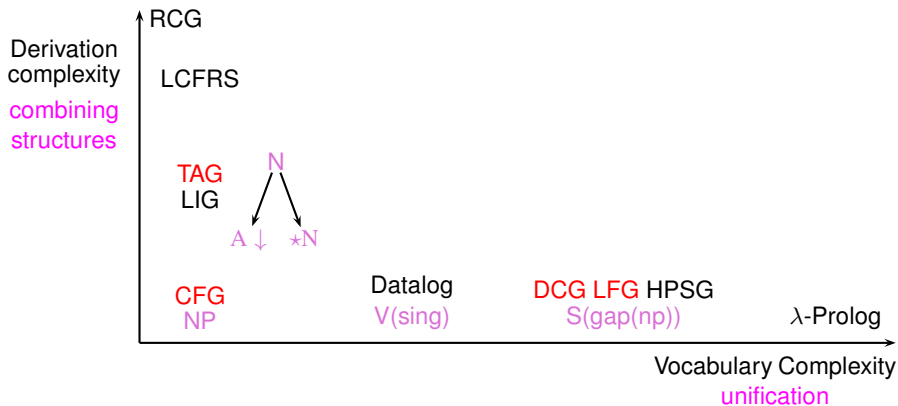
# Extending CFG with structures



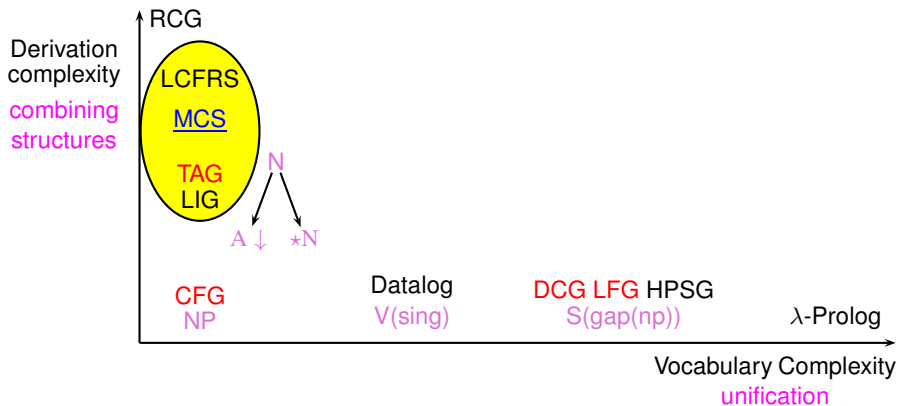
# Extending CFG with structures



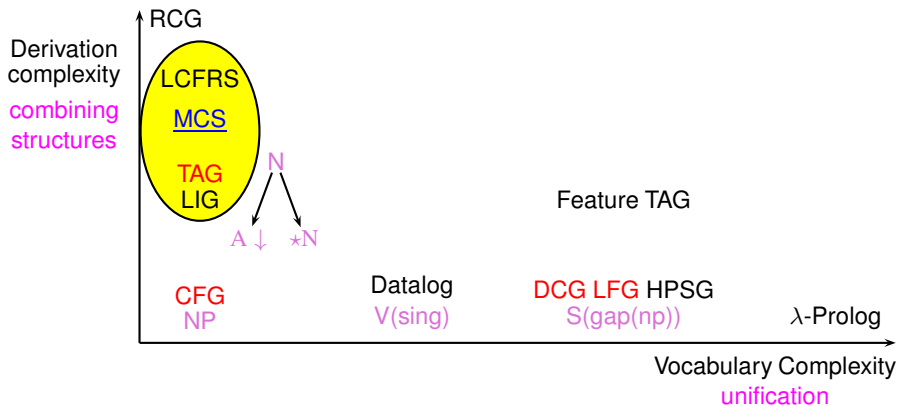
# Extending CFG with structures



# Extending CFG with structures



# Extending CFG with structures



- 1 Some background about TAGs
- 2 Deductive chart-based TAG parsing
- 3 Automata-based tabular TAG parsing



# From CFG to Tree Substitution Grammars

s → np vp  
np → pn  
np → det n  
np → np pp  
vp → v np  
vp → vp pp  
pp → prep  
np

CFG productions:

- are too local  
⇒ need decorations for info propagation
- are generally not *lexicalized*  
but info often propagated from words  
also more efficient parsing algo for lexicalized grammars

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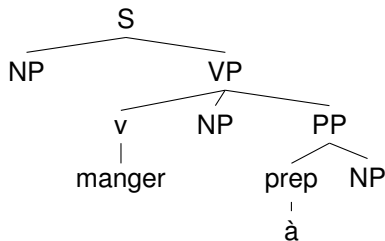
CFG productions:

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CFG productions can be grouped into trees

⇒ we get Tree Substitution Grammars (TSG)

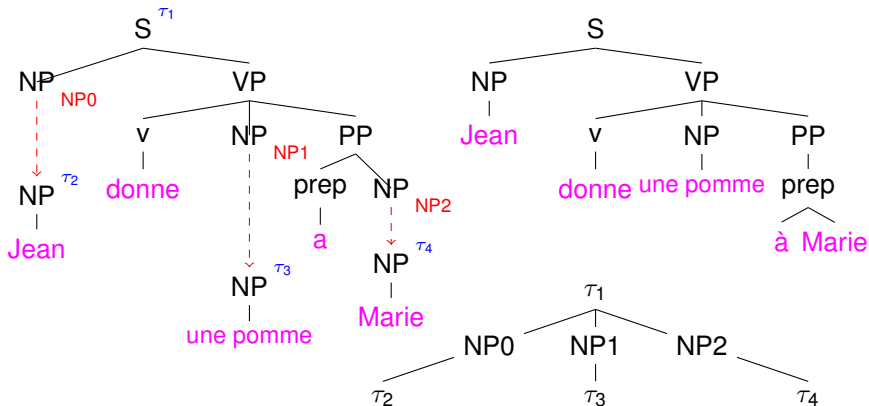
For instance, dealing with ditransitive  
verb **donner**



# Derivation Tree vs Parse Tree

TSG are strongly equivalent to CFG

However, for TSG, parse trees and derivation trees are not equivalent



Furthermore, several derivations may lead to a same parse tree

# One step farther: adjoining

How to deal with:

*Jean donne souvent une pomme à Marie*

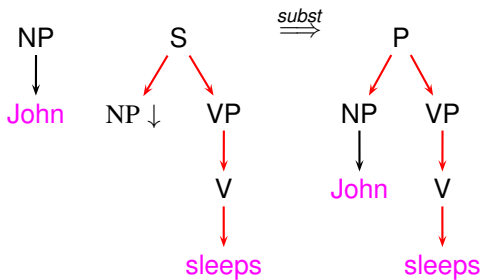
Need a way to insert the adverb somewhere in the verbal tree

⇒ adjoining operation

↪ Tree Adjoining Grammars

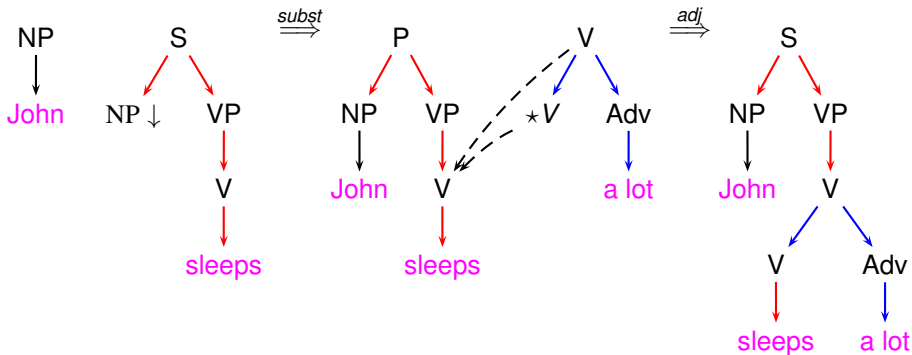
# TAG: a small example

Tree Adjoining Grammars [TAGs] [Joshi] build parse trees from initial and auxiliary trees by using 2 tree operations: substitution and **adjoining**

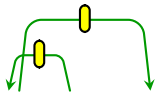


# TAG: a small example

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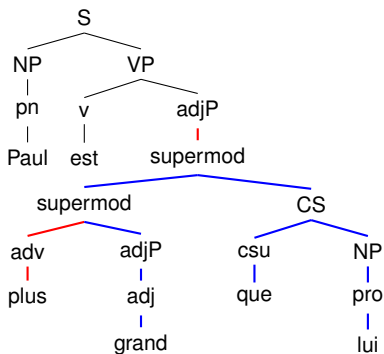
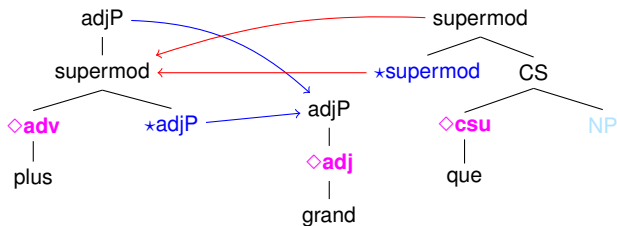


# A more complex example: French comparative



Paul est plus grand que lui

# A more complex example: French comparative





# TAGs: (First) Formal definition

A TAG  $G$  is a tuple  $(\mathcal{N}, \Sigma, S, \mathcal{I}, \mathcal{A})$  where

- $\Sigma$  a finite set of terminal symbols
- $\mathcal{N}$  a finite set of non-terminal symbols
- $S \in \mathcal{N}$  the axiom
- $\mathcal{I}$  and  $\mathcal{A}$  are two finite sets of **elementary trees** over  $\mathcal{N} \cup \Sigma \cup \{\epsilon\}$  only leaf nodes  $\nu$  may have a label  $l(\nu) \in \Sigma \cup \{\epsilon\}$ 
  - ▶ the trees  $\alpha$  in  $\mathcal{I}$  are **initial trees**
  - ▶ the trees  $\beta$  in  $\mathcal{A}$  are **auxiliary trees** and have a unique leaf node marked  $(\star)$  as a **foot**  $f_\beta$  with same label than the root node  $r_\beta$ , i.e.  $l(f_\beta) = l(r_\beta)$

Two operations may be used to combine the elementary trees

- substitution of a leaf node  $\nu$  of  $\gamma$  by some initial tree  $\alpha \in \mathcal{I}$ , ( $l(\nu) = l(t_\alpha)$ )
- adjoining of an (internal) node  $\nu$  of  $\gamma$  by some auxiliary tree  $\beta \in \mathcal{A}$

$G$  generates a tree language and a string language

$$T(G) = \{\gamma \mid \alpha \implies^* \gamma \wedge \text{yield}(\gamma) \in \Sigma^* \wedge \alpha \in \mathcal{I} \wedge r_\alpha = S\}$$

$$L(G) = \{\text{yield}(\gamma) \mid \gamma \in T(G)\}$$

Assuming  $\gamma = (V, E)$  with  $\nu \in V$  and  $\beta = (V_\beta, E_\beta)$  with  $r_\beta, f_\beta \in V_\beta$ , such that  $I(\nu) = I(r_\beta) \in \mathcal{N}$

$$\gamma[\text{adj}(\nu, \beta)] = (V', E')$$

with

$$\left\{ \begin{array}{l} V' = V \cup V_\beta \setminus \{\nu\} \\ E' = \cup \left\{ \begin{array}{l} \{(x, y) \in E \mid x \neq \nu \wedge y \neq \nu\} \\ E_\beta \\ \{(x, r_\beta) \mid (x, \nu) \in E\} \\ \{(f_\beta, y) \mid (\nu, y) \in E\} \end{array} \right. \end{array} \right.$$

**Note:** The node sets are assumed to be renamed to avoid clashes, i.e.  $E \cap E' = \emptyset$

A full definition of TAGs should include constraints on adjoining nodes:

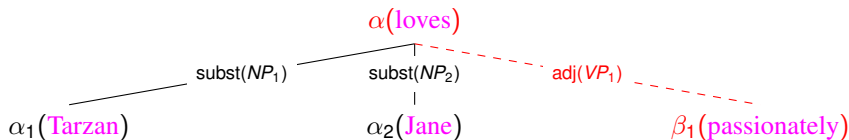
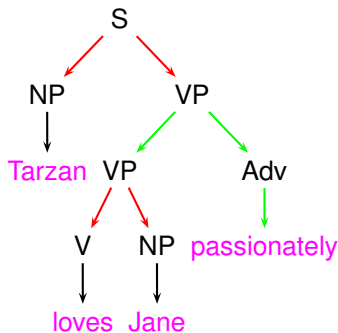
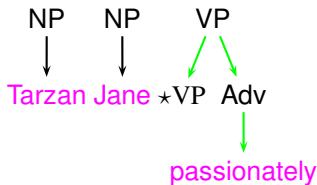
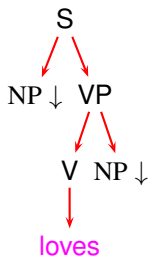
A TAG  $G$  is a tuple  $(\mathcal{N}, \Sigma, \mathcal{S}, \mathcal{I}, \mathcal{A}, f_{OA}, f_{SA})$  where, assuming  $V$  set of nodes in  $\mathcal{I} \cup \mathcal{A}$ ,

- $f_{OA} : V \mapsto \{0, 1\}$  specify if adjoining on  $\nu$  is obligatory (1) or not (0)
- $f_{SA} : V \mapsto 2^{\mathcal{A}}$  specify which auxiliary trees may be adjoined on  $\nu$   
**note:**  $\nu$  becomes non-adjoinable with  $f_{SA}(\nu) = \emptyset$

Adjoining constraints necessary for getting the full expressive power of TAGs but they are often implicit:

- no adjoining on leaf nodes (including foot nodes)
- explicit mandatory adjoining (MA, +) marks on some nodes
- explicit non adjoining (NA, -) marks on some nodes

# Derivation tree



For TAGs, **derivation tree** not isomorphic to parse tree but close from semantic level.

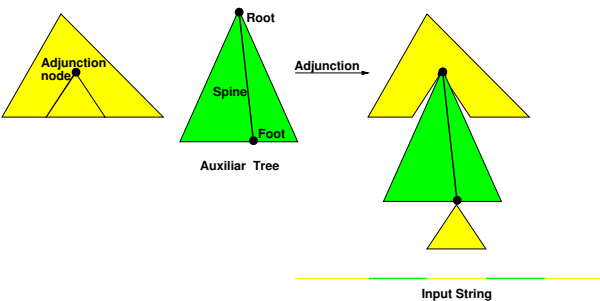
For a TAG  $G$ , its set of derivation trees  $D(G)$  forms a **regular tree language**

i.e.,  $D(G)$  may be generated by a finite tree automaton  
(top-down) term rewrite rules of the form

$$q_0 \leftarrow a(q_1, \dots, q_n), \quad q_i \in \mathcal{Q}, a \in \mathcal{F}$$

may also be seen as the parse trees for some CFG  $G'$

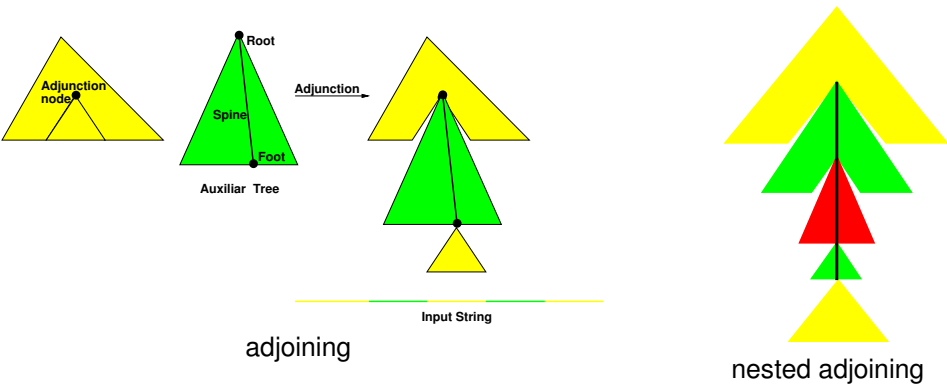
# TAG complexity: Adjoining



adjoining

- discontinuity (hole in aux tree)
- crossing (both sides of the hole)

# TAG complexity: Adjoining

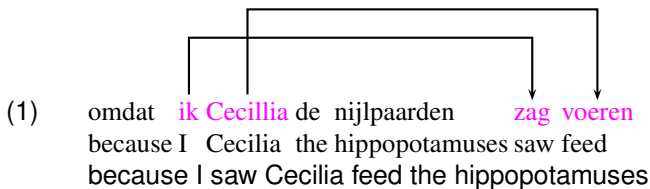


- discontinuity (hole in aux tree)
- crossing (both sides of the hole)
- unbounded synchronization (both sides of spine)

# Expressive power of TAGs

The adjoining operation extends the expressive power of TAGs w.r.t. CFGs.

- long distance dependencies (wh-pronoun extraction for instance)
- crossed dependencies as given by copy language “**ww**” or by language “ $a^n b^n c^n$ ”





TAGs can't handle the following languages:

- $a^n b^m c^n d^m e^n f^m$
- multiple copy languages  $w^n$  with  $n > 2$ .

Tree Adjoining Languages satisfy a **pumping lemma**

If  $L$  is a TAL, then there exists  $N$ , such for all  $w \in L$  and  $|w| > N$ , there exist  $x, y, z, v_1, v_2, w_1, w_2, w_3, w_4 \in \Sigma^*$ , such that

$$\begin{cases} |v_1 v_2 w_1 w_2 w_3 w_4| \leq N \\ |w_1 w_2 w_3 w_4| \geq 1 \end{cases}$$

and one of the following case holds

- 1  $w = xw_1 v_1 w_2 y w_3 v_2 w_4 z$  and  $\forall k \geq 0, xw_1^k v_1 w_2^k y w_3^k v_2 w_4^k z \in L$
- 2  $w = xw_1 v_1 w_2 v_2 w_3 y w_4 z$  and  $\forall k \geq 0, xw_1^{k+1} v_1 w_2 v_2 w_3 (w_2 w_4 w_3)^k y w_4 z \in L$
- 3  $w = xw_1 y w_2 v_1 w_3 v_2 w_4 z$  and *forall*  $k \geq 0, xw_1 y (w_2 w_1 w_3)^k w_2 v_1 w_3 v_2 w_4^{k+1} \in L$

# Closure properties of TALs

As CFLs, TALs form an Abstract Family of Languages (AFL):

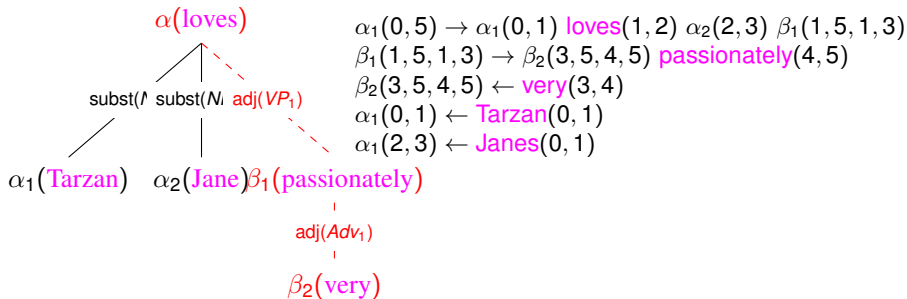
- 1 closed by intersection with regular languages
- 2 closed by union, concatenation, and Kleene-iteration
- 3 closed by homomorphism and inverse homomorphism

In particular, (1)  $\implies$  notion of **Shared Derivation Forest**

# Shared Derivation Forests

Formal definition in Vijay-Shanker & Weir 1993

0 *Tarzan* 1 *loves* 2 *Jane* 3 *very* 4 *passionately* 5



More formally, use tree nodes rather than trees

Space complexity in  $O(n^6)$  by binarization (adj on spine node  $\nu$ )

$$\nu^\top(i, j, r, s) \rightarrow r_\beta^\top(i, j, p, q) \nu^\perp(p, q, r, s)$$

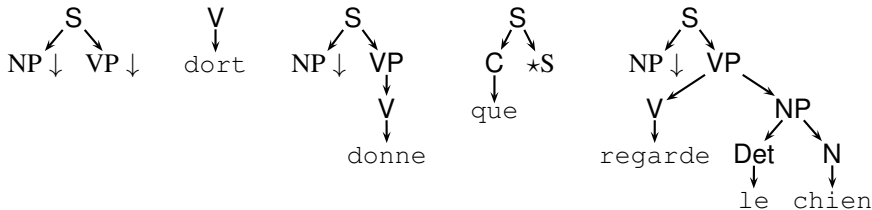
# Well formed trees

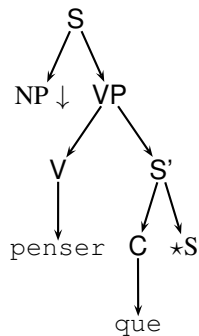
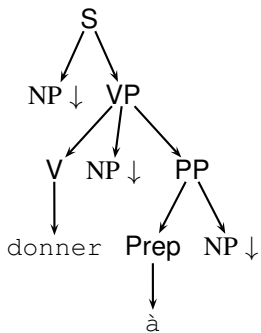
Many possible ways to define elementary trees

In practice, elementary trees follow some linguistic principles:

- lexical anchoring: at least, one non-empty lexical (frontier) node the head (or **anchor**)
- sub-categorization: a frontier node for each argument sub-categorized by the head  
**domain of locality**
- semantic consistency: a tree correspond to the scope of a semantic predicate with its arguments
- non-composition: a tree stands for a single semantic unit

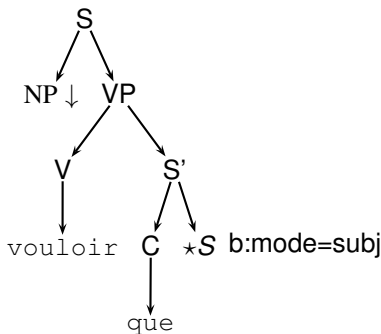
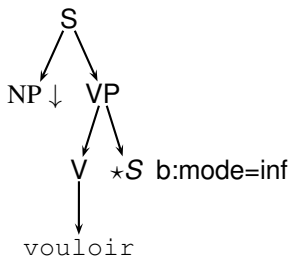
A few bad trees:





# Feature TAGs

The nodes may be decorated with a pair  $(t_{op}, b_{ot})$  of decorations

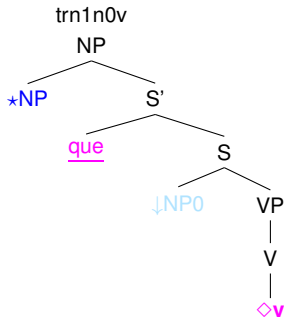
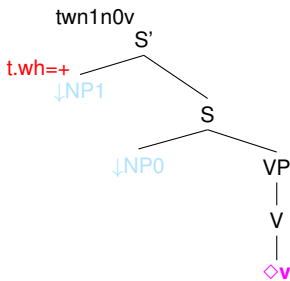
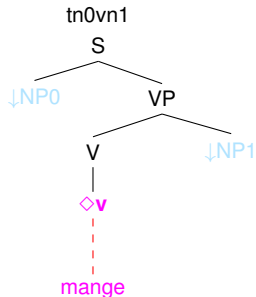


When adj on  $\nu$ , unification of  $\nu.top$  with  $r_{\beta}.top$  and  $\nu.bot$  with  $f_{\beta}.bot$   
alternate way to express adjoining constraints

**Note:** for flat decorations, same expressive power and complexity

# TAG families

Trees derived from a canonical ones grouped into **families**  
e.g. family of transitive verbs



+ all other extractions (on  $NP_0$ ) + passive + extractions on passive  
+ ordering + multiple **realizations** + ...

↪ XTAG architecture

- a set of trees (with **anchor nodes**) grouped into families
- a lexicon  $\mathcal{L}$  specifying for each word  $w$  the set of families it may anchor  
+ additional constraints



Large coverage TAG  $\implies$  many trees to write and maintain !

Alternative: generate the trees from a higher description level: [meta-grammars](#)

[Abeillé](#), [Candito](#)

- hierarchy of classes, containing constraints  
A precedes B, A dominates B, ...
- a class deals with a linguistic facet  
e.g. verb argument, refined into subject or object
- a class may require or provide functionalities
- the classes may be combined to form neutral classes
- the constraints of the neutral classes used to generate elementary trees

$\implies$  used for [FRMG](#), a large-coverage French TAG

<http://alpage.inria.fr/frmgwiki>

(plus mechanisms for factorizing elementary trees)

# Long-distance dependencies

(Recursive) Adjoining may replace LFG's functional uncertainty for long-distance dependencies

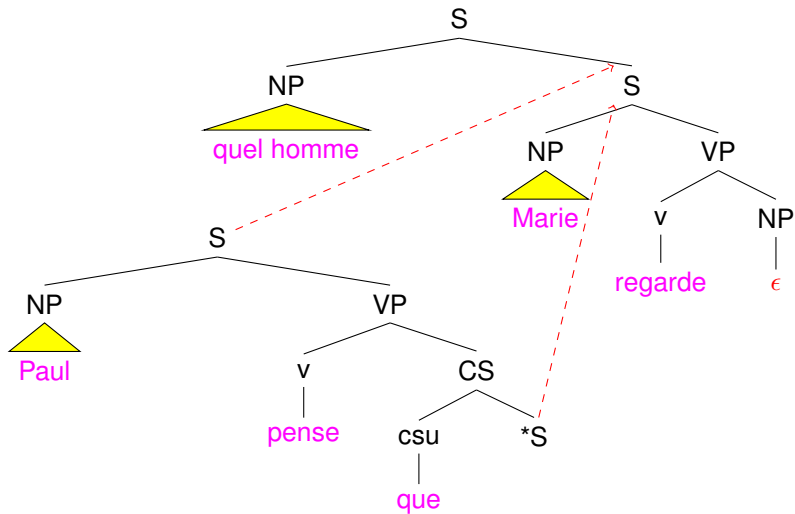
Jean demande [quel homme Paul pense [que Marie regarde  $\epsilon$ ]]



$S'$	$\longrightarrow$	NP		S
		$(\downarrow Wh) =_c +$		$\uparrow = \downarrow$
		$(\uparrow Focus) = \uparrow$		$(\downarrow Wh) = +$
		$(\uparrow Focus) = \uparrow (Comp)^* Obj$		

# Long-distance dependencies (TAGs)

Handled through repeated adjoining



- 1 Some background about TAGs
- 2 Deductive chart-based TAG parsing
- 3 Automata-based tabular TAG parsing

## Formalization of chart parsing

### Use of

- universe of tabulable **items**, representing (set of) partial parses
- items often build upon **dotted rules**
- chart edges labeled by dotted rules ( items  $\equiv \langle i, j, A \leftarrow \alpha \bullet \beta \rangle$  )
- a **deductive system** specifying how to derive items

$$A_0 \leftarrow A_1 \dots A_i \bullet A_{i+1} \dots A_n$$

# CKY as a deductive system (for CFGs)

$$\frac{}{\langle i, i, A \leftarrow \bullet \alpha \rangle} \quad \exists A \leftarrow \alpha \quad \begin{array}{c} A \leftarrow \bullet \alpha \\ \text{---} \\ i \end{array} \quad \text{(Seed)}$$

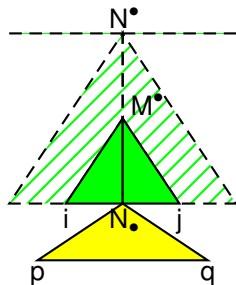
$$\frac{\langle i, j, A \leftarrow \alpha \bullet a \beta \rangle}{\langle i, j+1, A \leftarrow \alpha a \bullet \beta \rangle} \quad a = a_{j+1} \quad \begin{array}{c} \text{---} \\ \text{---} \\ i \quad \xrightarrow{A \leftarrow \alpha a \bullet \beta} \quad j \quad \xrightarrow{A \leftarrow \alpha \bullet a \beta} \quad j+1 \end{array} \quad \text{(Scan)}$$

$$\frac{\langle i, j, A \leftarrow \alpha \bullet B \beta \rangle \quad \langle j, k, B \leftarrow \gamma \bullet \rangle}{\langle i, k, A \leftarrow \alpha B \bullet \beta \rangle} \quad \begin{array}{c} \text{---} \\ \text{---} \\ i \quad \xrightarrow{A \leftarrow \alpha \bullet B \beta} \quad j \quad \xrightarrow{B \leftarrow \gamma \bullet} \quad k \end{array} \quad \text{(Complete)}$$

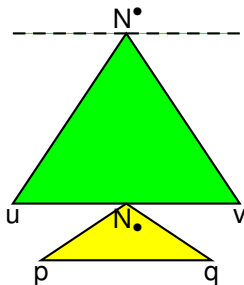
## CKY algorithm for TAGs [Vijay-Shanker & Joshi 85]

Presentation:

- **Dotted trees**  $N^\bullet$  and  $N_\bullet$ , where  $N$  is a node of an elementary tree
- Items  $\langle N^\bullet, i, p, q, j \rangle$  and  $\langle N_\bullet, i, p, q, j \rangle$  with  $p, q$  possibly covering a foot node.



$\langle M_\bullet, i, p, q, j \rangle$



Without adjoining:  $\langle N_\bullet, p, -, -, q \rangle$

With adjoining:  $\langle N^\bullet, u, -, -, v \rangle$

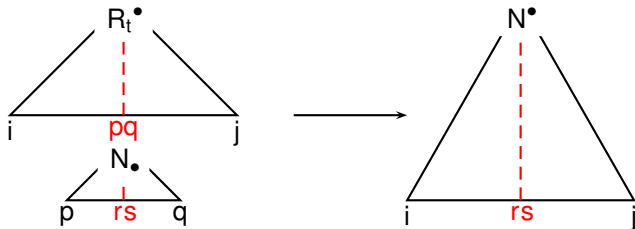
# Rule (Adjoin)

Gluing a sub-tree at a foot node.

$$\frac{\langle N_{\bullet}, p, r, s, q \rangle \langle R_t^{\bullet}, i, p, q, j \rangle}{\langle N^{\bullet}, i, r, s, j \rangle}$$

$$\text{label}(N) = \text{label}(R_t)$$

(Adjoin)





When no adjoining on a node

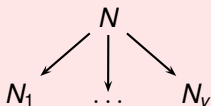
$$\frac{\langle N_{\bullet}, p, r, s, q \rangle}{\langle N^{\bullet}, p, r, s, q \rangle}$$

(NoAdjoin)

# Rule (Complete)

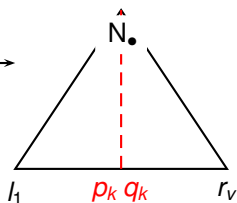
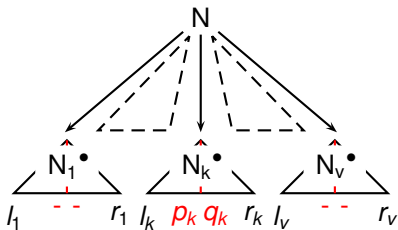
## Gluing all node's children

$$\frac{\langle N_i^\bullet, l_i, p_i, q_i, r_i \rangle_{i=1, \dots, v}}{\langle N_\bullet, l_1, \cup p_i, \cup q_i, r_v \rangle}$$



and  $\forall i, l_{i+1} = r_i$  (Complete)

**Note:** At most one child ( $k$ ) covers a foot node with  $(\cup p_i, \cup q_i) = (p_k, q_k)$



Other deductive rules needed to handle

- 1 substitution
- 2 terminal scanning

+ axioms

Time complexity  $O(n^{\max(6, 1+v+2)})$  with

- $v$  : maximal number of children per node
- 2 : number of indexes to cover a possible unique foot node

Normalization using **binary-branching trees** ( $v = 2$ )  $\implies$  complexity  $O(n^6)$

4 indexes per item  $\implies$  Space complexity in  $O(n^4)$  for a recognizer  
 $O(n^6)$  for a parser, keeping **backpointers** to parents

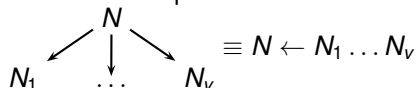
Optimal worst-case complexities

but practically, even less efficient than CKY for CFGs

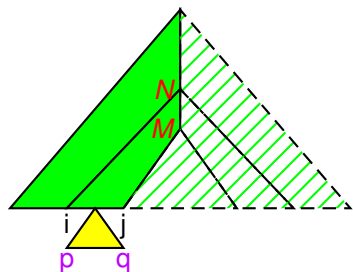
# Prediction, dotted trees and dotted productions

To mark prediction, new dotted trees [Shabes]:  $\bullet N$  and  $\cdot N$

Alternative: equivalence with dotted productions



dotted tree	dotted production
$N_k^\bullet, \bullet N_{k+1}$	$N \leftarrow N_1 \dots N_k \bullet N_{k+1} \dots N_v$
$\bullet R$ (root)	$\top \leftarrow \bullet R$
$R^\bullet$ (root)	$\top \leftarrow R^\bullet$
$\cdot N$	$N \leftarrow \bullet N_1 \dots N_v$
$N_\bullet$	$N \leftarrow N_1 \dots N_n \bullet$



$$\langle N \leftarrow \alpha \bullet M \beta, i, p, q, j \rangle$$

- Glue a sub-tree at foot node  $F_t$  (maybe useless !)

$$\frac{\langle M \leftarrow \gamma \bullet, p, r, s, q \rangle \quad \langle T \leftarrow R_t \bullet, i, p, q, j \rangle}{\langle M \leftarrow \gamma \bullet, i, r, s, j \rangle} \quad \text{label}(M) = \text{label}(R_t) \quad (\text{Adjoin})$$

- Advance in recognition of  $N$ 's children

$$\frac{\langle N \leftarrow \alpha \bullet M \beta, i, u, v, j \rangle \quad \langle M \leftarrow \gamma \bullet, j, r, s, k \rangle}{\langle N \leftarrow \alpha M \bullet \beta, i, u \cup r, v \cup s, k \rangle} \quad (\text{Complete})$$

(Adjoin) and (Complete) similar to CKY (binary form)

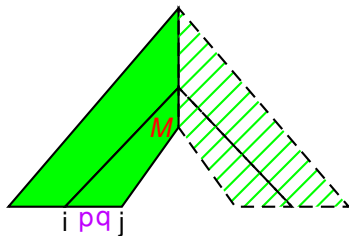
# Adjoining Prediction

Predict adjoining at  $M$

$$\langle \underline{N \leftarrow \alpha \bullet M \beta}, i, p, q, j \rangle$$

$$\text{label}(M) = \text{label}(R_t)$$

(CallAdj)



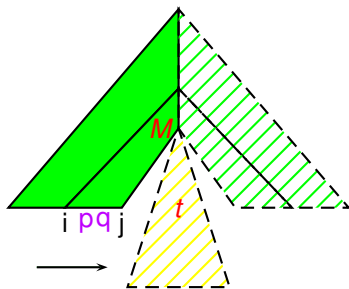
# Adjoining Prediction

Predict adjoining at  $M$

$$\frac{\langle N \leftarrow \alpha \bullet M \beta, i, p, q, j \rangle}{\langle T \leftarrow \bullet R_t, j, -, -, j \rangle}$$

$$\text{label}(M) = \text{label}(R_t)$$

(CallAdj)



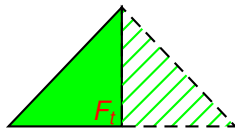
# Foot Prediction

Predict a sub-tree root at  $M$  to recognize below foot node  $F_t$

$$\langle \underline{F_t \leftarrow \bullet \perp, i, -, -, i} \rangle$$

$$\text{label}(F_t) = \text{label}(M)$$

(CallFoot)





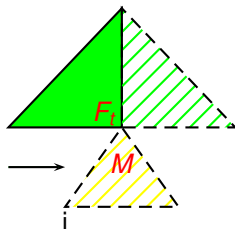
# Foot Prediction

Predict a sub-tree root at  $M$  to recognize below foot node  $F_t$

$$\frac{\langle F_t \leftarrow \bullet \perp, i, -, -, i \rangle}{\langle M \leftarrow \bullet \gamma, i, -, -, i \rangle}$$

$$\text{label}(F_t) = \text{label}(M)$$

(CallFoot)



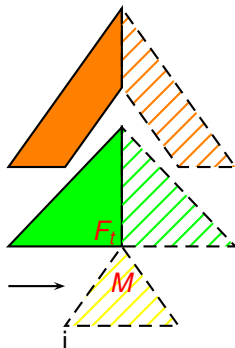
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Predict a sub-tree root at  $M$  to recognize below foot node  $F_t$

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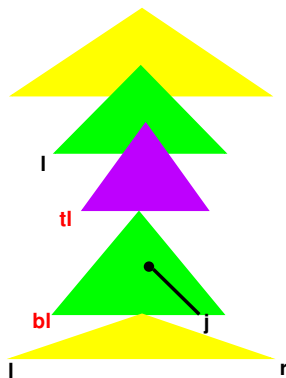
(CallFoot)



The prediction of  $M$  not related to the node  $M'$  having triggered the adjoining of  $t$   
 $\Rightarrow$  Non prefix valid parsing strategy

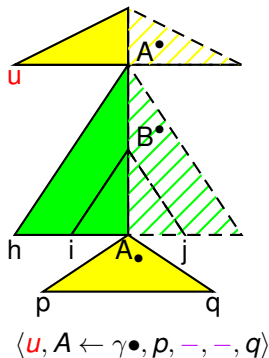
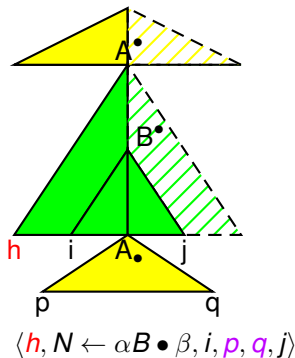
- Space complexity remains  $O(n^4)$
- Dotted productions  $\implies$  implicit binarization  $\implies$  time in  $O(n^6)$
- Non prefix valid: impact difficult to evaluate in practice
- **Note:** Dotted productions also applicable to improve CKY

Complexities time in  $O(n^9)$  and space in  $O(n^6)$  due to 6-index items



Actually, *tl* and *bl* may be avoided using dotted productions

Item with only an extra index  $h$ :  $\langle h, N \leftarrow \alpha \bullet \beta, i, p, q, j \rangle$   
 $h$  states starting (leftmost) position of current tree

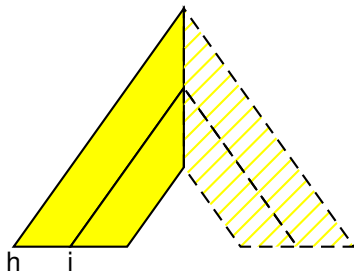


$$\langle h, N \leftarrow \alpha \bullet M^{\beta, i, p, q, j} \rangle$$

---

$$\text{label}(F_t) = \text{label}(M)$$

(CallFootPf)



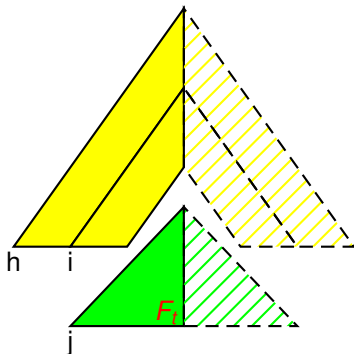
$$\langle h, N \leftarrow \alpha \bullet M \beta, i, p, q, j \rangle$$

---

$$\langle j, F_t \leftarrow \bullet \perp, k, -, -, k \rangle$$

$$\text{label}(F_t) = \text{label}(M)$$

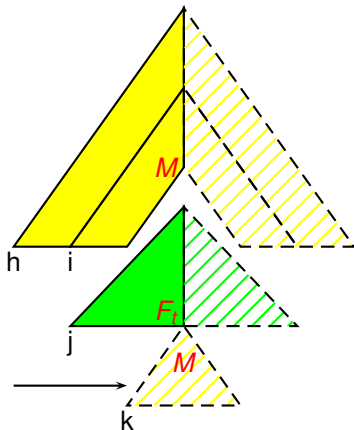
(CallFootPf)



$$\frac{\langle h, N \leftarrow \alpha \bullet M \beta, i, p, q, j \rangle}{\langle h, M \leftarrow \bullet \gamma, k, -, -, k \rangle}$$

$$\text{label}(F_t) = \text{label}(M)$$

(CallFootPf)





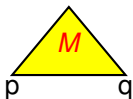
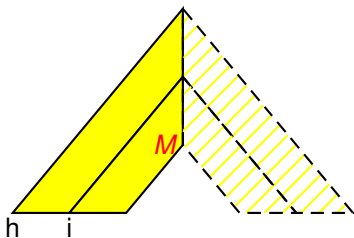
# Adjoining return

$$\langle h, N \leftarrow \alpha \bullet M \beta, i, u, v, j \rangle$$

$$\langle h, M \leftarrow \gamma \bullet, p, r, s, q \rangle$$

---

$$\text{label}(M) = \text{label}(R_t) \quad (\text{AdjoinPf})$$

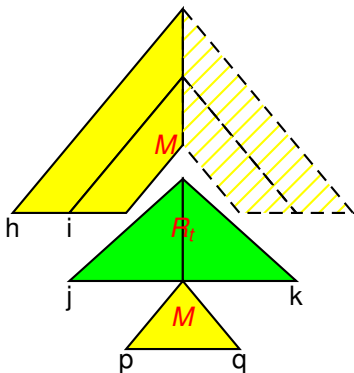


# Adjoining return

$$\begin{aligned} \langle h, N \leftarrow \alpha \bullet M \beta, i, u, v, j \rangle \\ \langle j, \top \leftarrow R_t \bullet, j, p, q, k \rangle \\ \langle h, M \leftarrow \gamma \bullet, p, r, s, q \rangle \end{aligned}$$

---

$$\text{label}(M) = \text{label}(R_t) \quad (\text{AdjoinPf})$$



# Adjoining return

$$\langle h, N \leftarrow \alpha \bullet M \beta, i, u, v, j \rangle$$

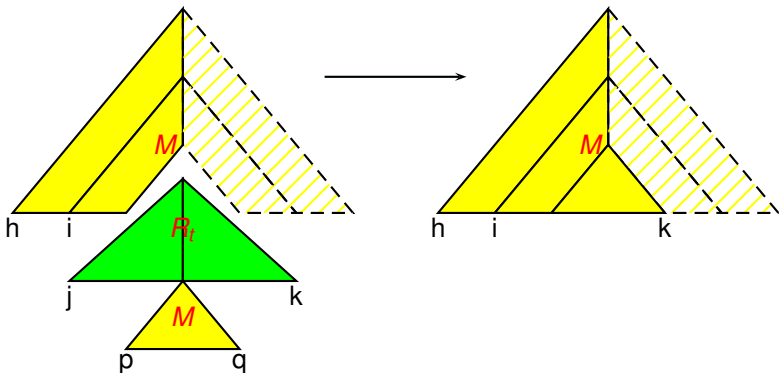
$$\langle j, \top \leftarrow R_t \bullet, j, p, q, k \rangle$$

$$\langle h, M \leftarrow \gamma \bullet, p, r, s, q \rangle$$

---

$$\langle h, N \leftarrow \alpha M \bullet \beta, i, u \cup r, v \cup s, k \rangle$$

$$\text{label}(M) = \text{label}(R_t) \quad (\text{AdjoinPf})$$



Maximal time complexity provided by (AdjoinPf) :  $O(n^{10})$  because of 10 indexes

$$\frac{\begin{array}{l} \langle h, N \leftarrow \alpha \bullet M \beta, i, u, v, j \rangle \\ \langle j, \top \leftarrow R_t \bullet, j, p, q, k \rangle \\ \langle h, M \leftarrow \gamma \bullet, p, r, s, q \rangle \end{array}}{\langle h, N \leftarrow \alpha M \bullet \beta, i, u \cup r, v \cup s, k \rangle} \quad \text{label}(M) = \text{label}(R_t) \quad (\text{AdjoinPf})$$

But  $(u, v)$  or  $(r, s)$  equals  $(-, -)$

$\implies$  (Case analysis) splitting rule into 2 sub-rules  $\implies O(n^8) \implies$  not sufficient !

# Splitting and intermediary structures

Split (AdjoinPf) into 2 successive steps with an intermediary structure

$$[M \leftarrow \gamma \bullet, j, r, s, k]$$

This intermediary structure combines the aux. tree with the subtree rooted at  $M$

$$\frac{\langle j, \top \leftarrow R_t \bullet, j, p, q, k \rangle \quad \langle h, M \leftarrow \gamma \bullet, p, r, s, q \rangle}{[M \leftarrow \gamma \bullet, j, r, s, k]} \quad (\text{AdjoinPf-1})$$

$$\frac{\langle h, N \leftarrow \alpha \bullet M \beta, i, u, v, j \rangle \quad [M \leftarrow \gamma \bullet, j, r, s, k] \quad \langle h, M \leftarrow \gamma \bullet, p, r, s, q \rangle}{\langle h, N \leftarrow \alpha M \bullet \beta, i, u \cup r, v \cup s, k \rangle} \quad (\text{AdjoinPf-2})$$

$$\frac{\langle j, \top \leftarrow R_{t\bullet}, j, p, q, k \rangle \quad \langle h, M \leftarrow \gamma_{\bullet}, p, r, s, q \rangle}{[M \leftarrow \gamma_{\bullet}, j, r, s, k]} \quad (\text{AdjoinPf-1})$$

Involves 7 indexes  $\{j, p, q, k, h, r, s\}$  but  $h$  not consulted

$$\frac{\langle h, M \leftarrow \gamma_{\bullet}, p, r, s, q \rangle}{\langle \star, M \leftarrow \gamma_{\bullet}, p, r, s, q \rangle} \quad (\text{Proj})$$

$$\frac{\langle j, \top \leftarrow R_{t\bullet}, j, p, q, k \rangle \quad \langle \star, M \leftarrow \gamma_{\bullet}, p, r, s, q \rangle}{[M \leftarrow \gamma_{\bullet}, j, r, s, k]} \quad (\text{AdjoinPf-1})$$

Finally,  $O(n^6)$  time complexity

# Case of (AdjoinPf-2)

$$\frac{\langle h, N \leftarrow \alpha \bullet M \beta, i, u, v, j \rangle \quad [M \leftarrow \gamma \bullet, j, r, s, k] \quad \langle h, M \leftarrow \gamma \bullet, p, r, s, q \rangle}{\langle h, N \leftarrow \alpha M \bullet \beta, i, u \cup r, v \cup s, k \rangle}$$

(AdjoinPf-2)

10 indexes  $\implies$  Raw complexity in  $O(n^{10})$

At least one pair in  $(u, v)$  or  $(r, s)$  equals  $(-, -)$ ;

Case splitting  $\implies O(n^8)$

Pair  $(p, q)$  not consulted; projection  $\implies O(n^6)$

Rule splitting, intermediary structures, and projections decrease complexities but increase the number of steps  
To be practically validated !

Designing a tabular algorithm for TAGs is complex!

- Designing items
- Understanding the invariants
- Formulating the deductive rules (simultaneously handling tabulation and strategy)
- Optimizing rules (splitting and projections)

How to adapt for close formalisms such as [Linear Indexed Grammars](#) [LIG]?

$$A_0([\circ \circ x]) \leftarrow A_1([\ ])\dots A_k([\circ \circ y])\dots A_n([\ ])$$



**Indexed Grammars:** Context-Free grammars with non terminals decorated with stacks

Linear Indexed Grammars: a single stack propagated per production

A LIG  $G = (\mathcal{N}, \Sigma, \mathcal{I}, S, \mathcal{P})$  where

- $\mathcal{I}$  is a finite set of indices
- $\mathcal{P}$  is a finite set of productions of the form

$$A[\alpha] \rightarrow A_1[] \dots A_i[\beta] \dots A_n[]$$

or

$$A[] \rightarrow \gamma$$

with  $\gamma \in \Sigma^*$  and  $\alpha, \beta \in \mathcal{I}^*$

Relationship with (linear monadic) Context-Free Tree Languages

LIGs and TAGs are weakly equivalent, and almost strongly equivalent

TAGs may be easily encoded by LIGs, using tree nodes as non-terminals

- adjoining node  $\nu$  in  $\gamma$  using aux. tree  $\beta$

$$\nu[\circ\circ] \rightarrow r_\beta[\circ \circ \nu]$$

- discharging a node  $\nu$  with children  $\nu_1, \dots, \nu_n$  at a foot node  $f_\beta$

$$f_\beta[\circ \circ \nu] \leftarrow \nu_1[\alpha_1] \dots \nu_n[\alpha_n]$$

where  $\alpha_i = [\circ\circ]$  if  $\nu_i$  on spine, and  $\alpha_i = []$  otherwise

- traversing a node  $\nu$  without adjoining

$$\nu[\circ\circ] \leftarrow \nu_1[\alpha_1] \dots \nu_n[\alpha_n]$$

with same conditions on  $\alpha_i$  than above

Reverse way more difficult: no locality constraint between push and pop points  
(same aux. tree  $\beta$  for TAGs)

Suggest using LPDAs to parse LIGs and TAGs  
but non efficient and non termination

- 1 Some background about TAGs
- 2 Deductive chart-based TAG parsing
- 3 Automata-based tabular TAG parsing

# From formalisms to automata

## Methodology:

- Automata are operational devices used to describe the steps of [Parsing Strategies](#)
- Dynamic Programming interpretations of automata used to identify [context-free subderivations](#) that may be tabulated.

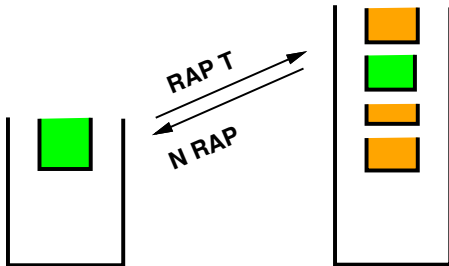
Formalisms	Automata	Tabulation	Notes
RegExp	FSA	-	
CFG	PDA	$O(n^3)$	Lang
TAG / LIG	2-Stack Automata Embedded PDA	$O(n^6)$ $O(n^6)$	Becker, Clergerie & Pardo Nederhof

**Problem:** 2-stack automata (or EPDA) have the power of Turing Machine (intuition) moving left- or rightward  $\equiv$  pushing on first or second stack & popping the other one

$\implies$  need restrictions

Embedded Push-Down Automata **Becker** are natural candidates for LIGs (and TAGs) by handling stack of stacks.

Two flavors: Top-Down and Bottom-Up EPDAs



# 2-stack automata for TAGs

Solution: stack asymmetry

**Master Stack:** to keep trace of uncompleted tree traversals

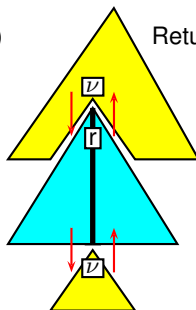
**Auxiliary Stack:** only to keep trace of uncompleted adjunctions

Adjunction info: (top-down)  $\bar{\nu}^n = \nu$  and (bottom-up)  $\underline{\nu}_n = \perp$

$\bullet T, T\bullet, \bullet B, B\bullet$ : prediction and propagation info about top and bottom node decorations (Feature TAGs)

Calls (top-down prediction)

Returns (bottom-up propagation)



# 2-stack automata for TAGs

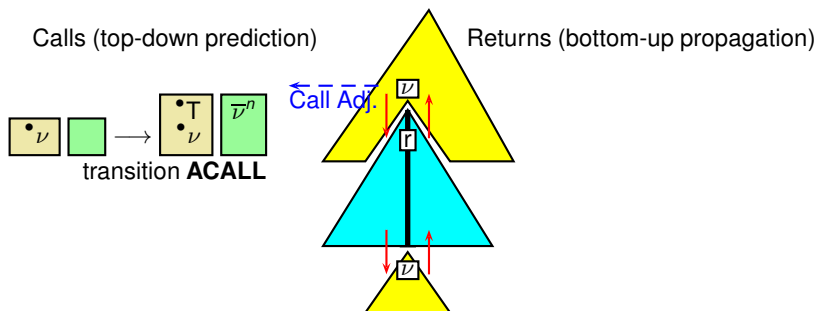
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# 2-stack automata for TAGs

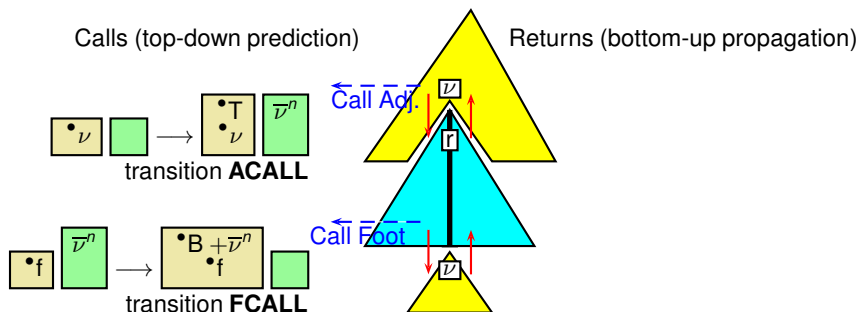
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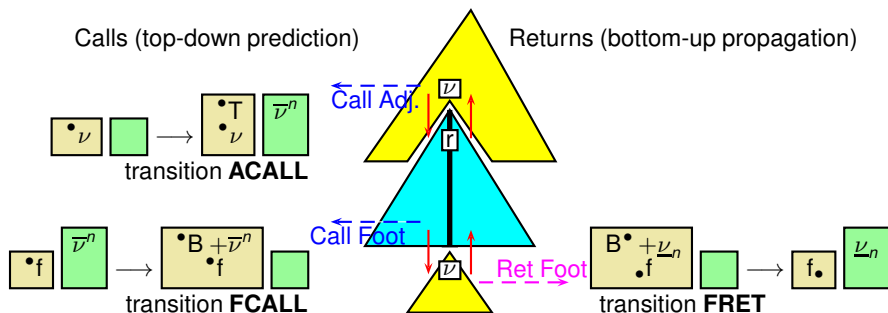
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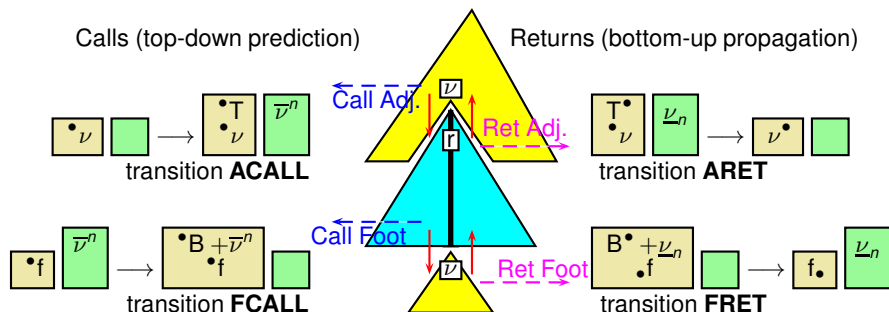
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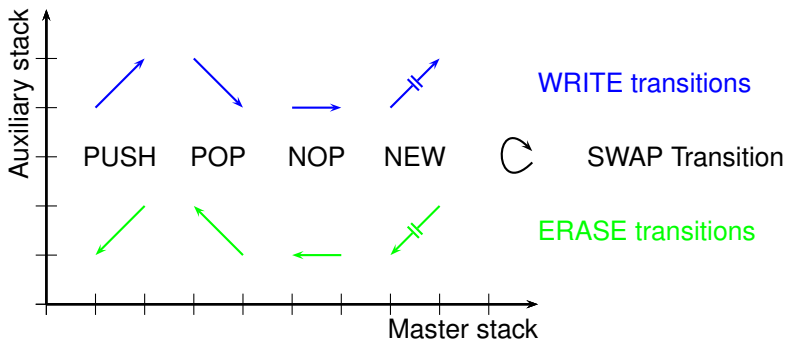


# Transitions

**Retracing** in **erase** mode concerns only the size of **AS** (not its content).

**Retracing** possible because :

*WRITE* transitions leave *marks* (*PUSH*, *POP*, *NOP*, *NEW*) in the Master Stack that can only be removed by a dual *ERASE* transition.

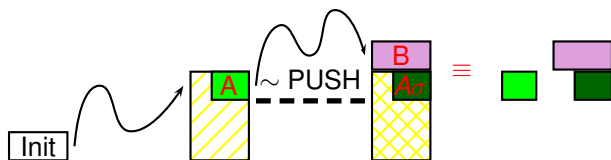


**Dynamic Programming** : Recursive decomposition of problems into elementary subproblems that may be **combined**, **tabulated**, and **reused**  
eg the knapsack problem

# Context-Free derivation for PDAs

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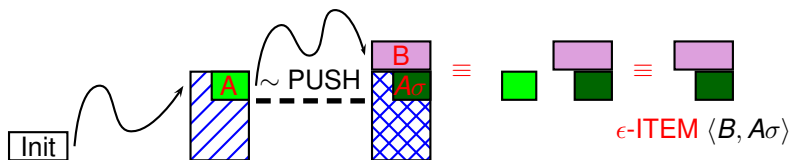
For PDAs, derivations broken into elementary **Context-Free sub-derivations**:



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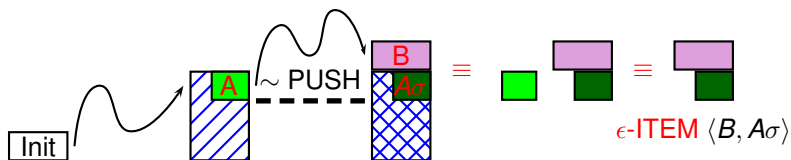
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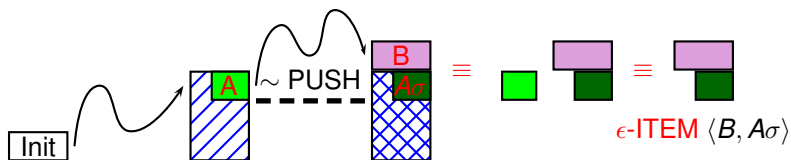
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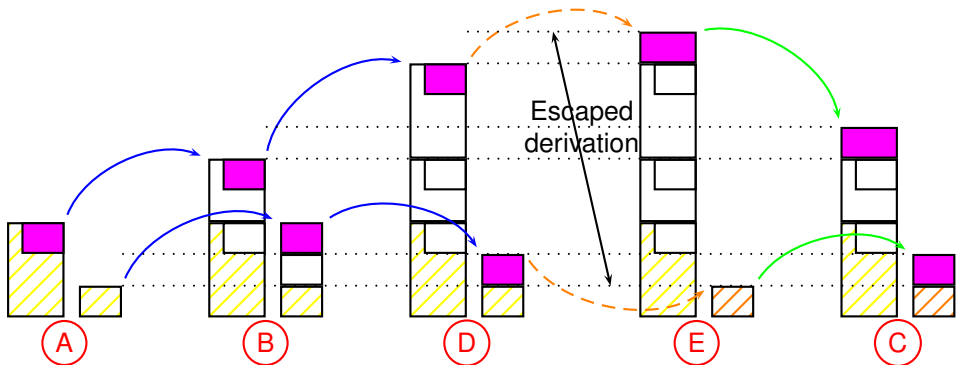
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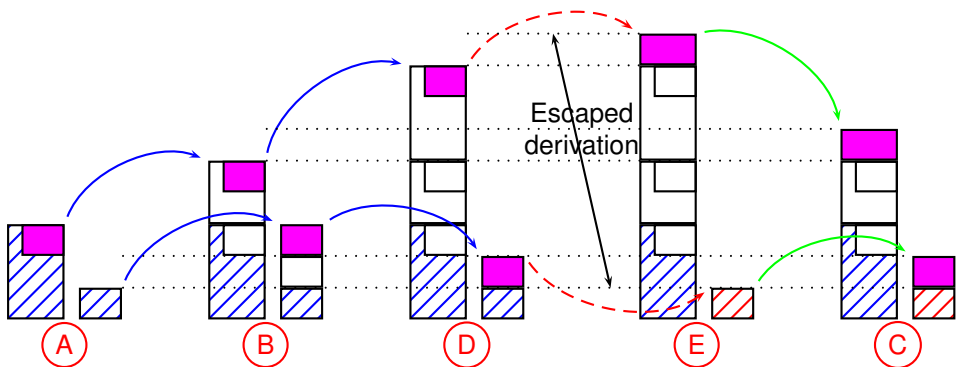
**A** is the fraction  $\epsilon$  of information consulted to trigger the subderivation and not propagated to **B**.



# (Escaped) CF derivations for 2SA



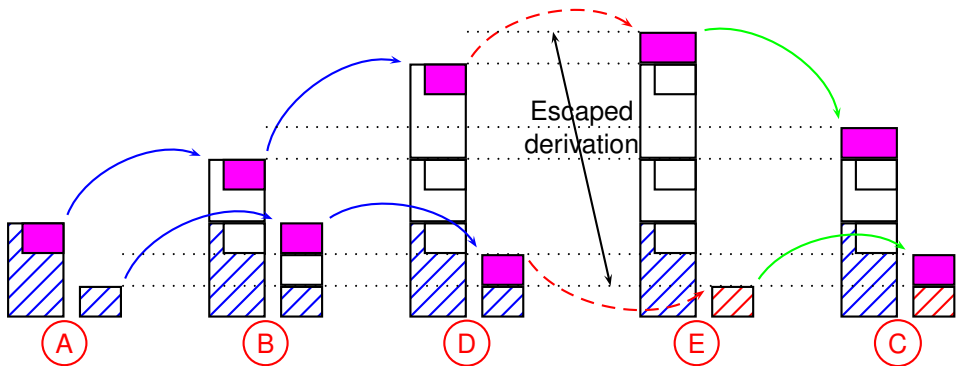
# (Escaped) CF derivations for 2SA



$\Rightarrow$  5-point xCF items  $AB[DE]C = \langle \epsilon A \rangle \langle \epsilon B, b \rangle [ \langle \epsilon D, d \rangle \langle E \rangle ] \langle C, c \rangle$   
 $[TAG] \rightsquigarrow \langle \epsilon A \rangle \langle \epsilon B \rangle [ \langle \epsilon D \rangle \langle E \rangle ] \langle C \rangle$

When no escaped part  $\Rightarrow$  3-point CF items  $ABC = \langle \epsilon A \rangle \langle \epsilon B, b \rangle \langle C \rangle$

(new generalization) escaped part  $[DE]$  may take place between  $A$  and  $B$

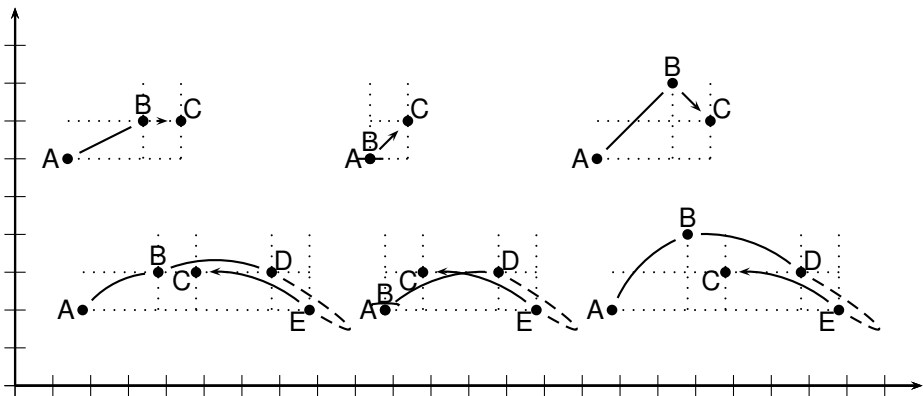


- *A* root of elementary tree
- *B* start of adjoining
- *C* current position in the tree
- *D* and *E* left and right borders of the foot

# Item shapes

At most 5 indexes per items  $\implies$  Space complexity in  $O(n^5)$

SD-2SA restrictions & transition kinds  $\implies$  6 possible item shapes



# Combining items and transitions

By graphically playing with items and transitions, we find 10 composition rules with  $O(n^8)$  time complexity

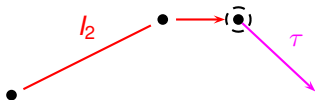
may be split into 11 rules with  $O(n^6)$  time complexity

(Easy:) Write a POP mark:  $l_1 + l_2 + \tau = l_3$

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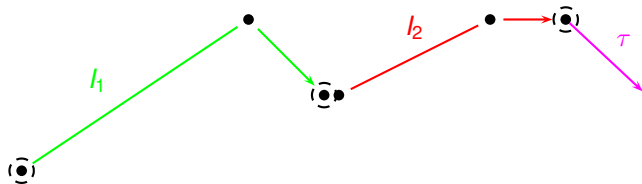


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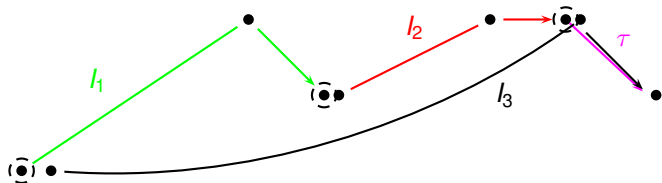


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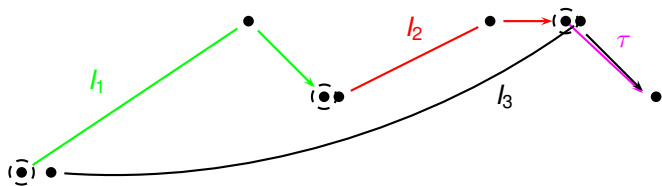




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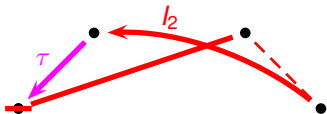
Consultation of 3 indexes  $(\odot) \implies$  Complexity  $O(n^3)$

## Combining items and transitions (2)

(complex:) Erasing a PUSH mark:  $l_1 + l_2 + l_3 + \tau = l_4$   
e.g. when returning from auxiliary tree (ending adjoining)

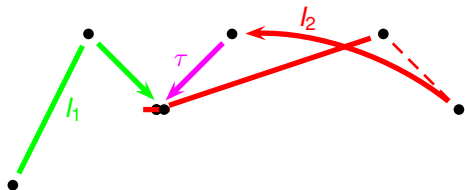
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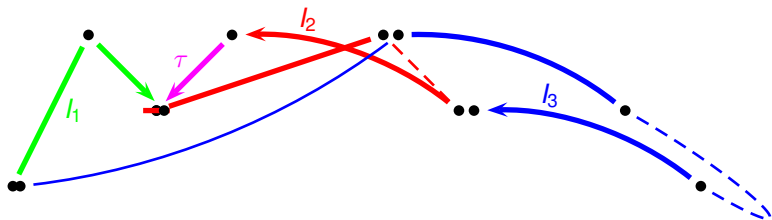
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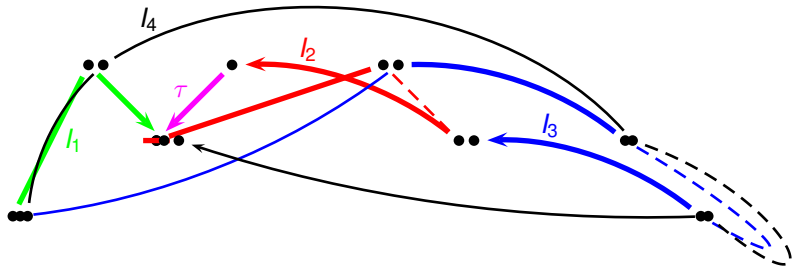
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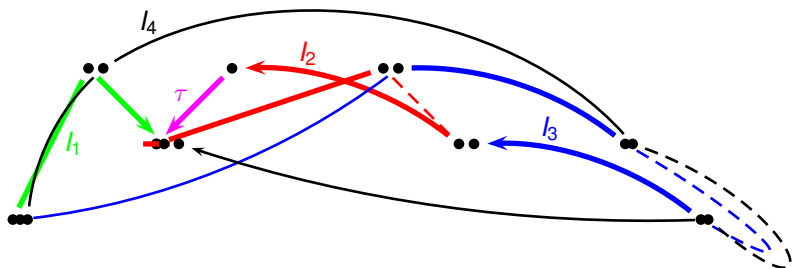
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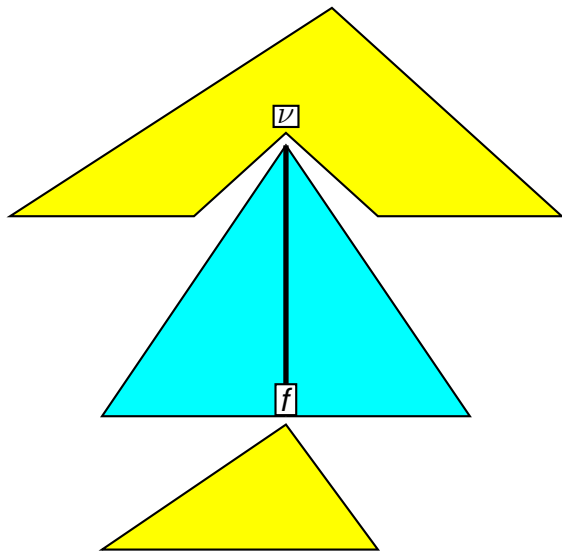
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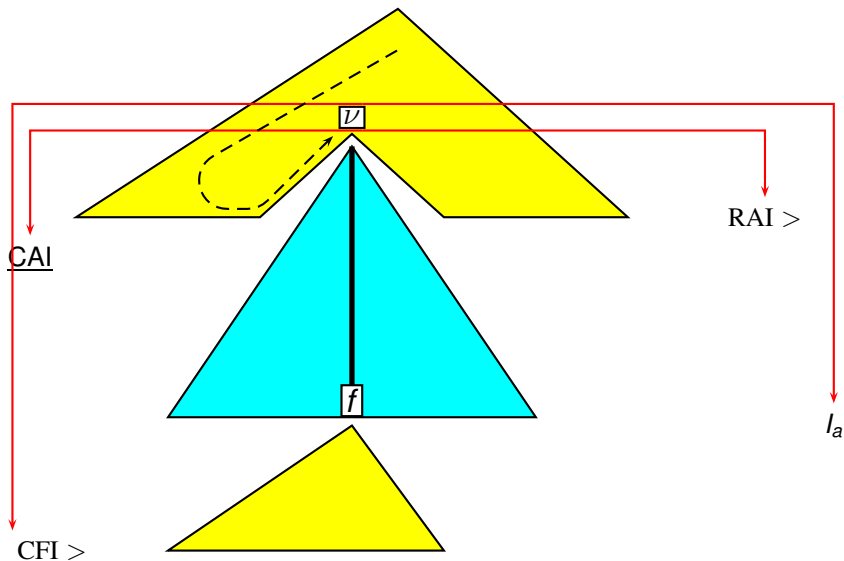
- Consultation of 8 indexes ( $\odot$ )  $\implies$  Complexity  $O(n^8)$
- need to decompose, project and use intermediary steps (as seen before)

# Cascade of partial evaluations

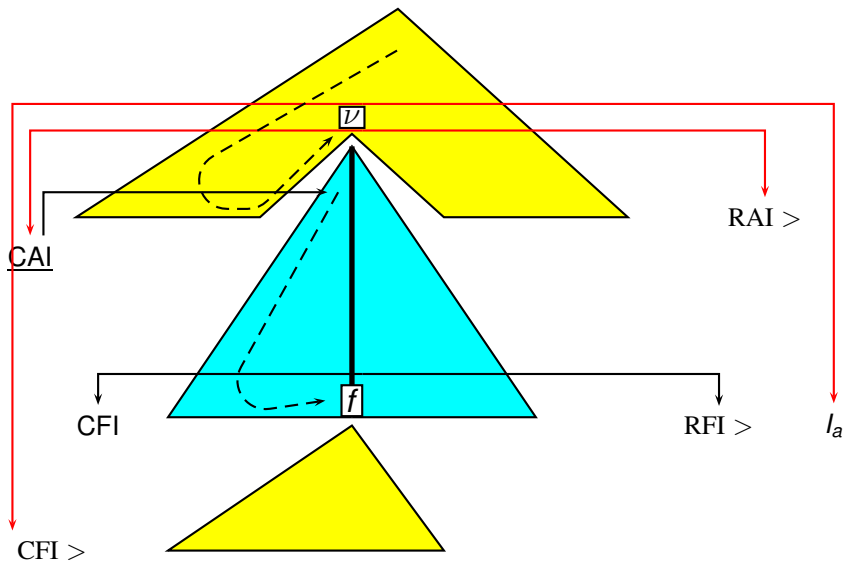




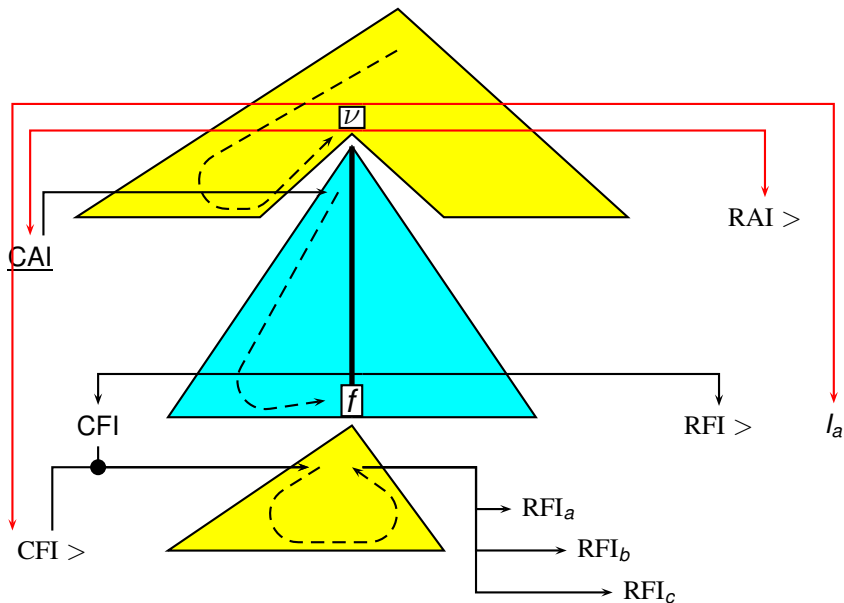
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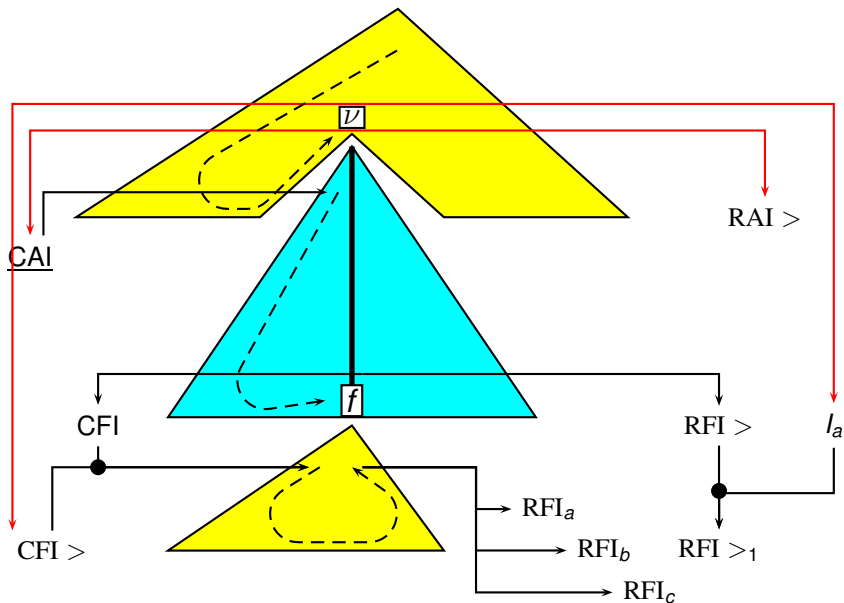
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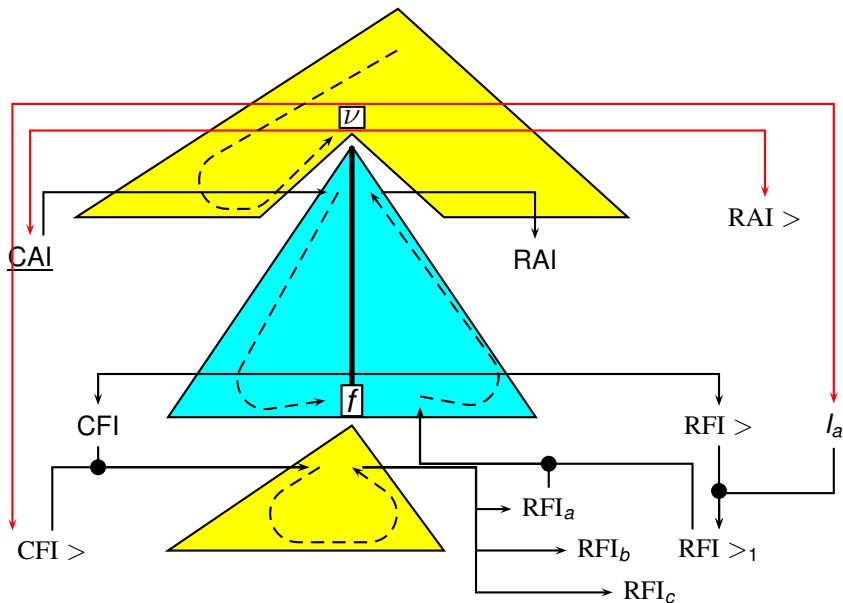
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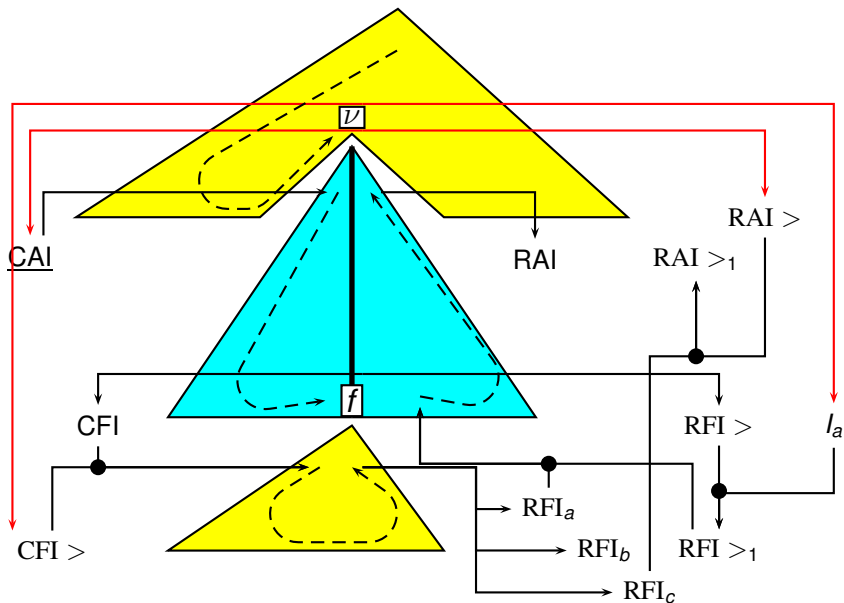
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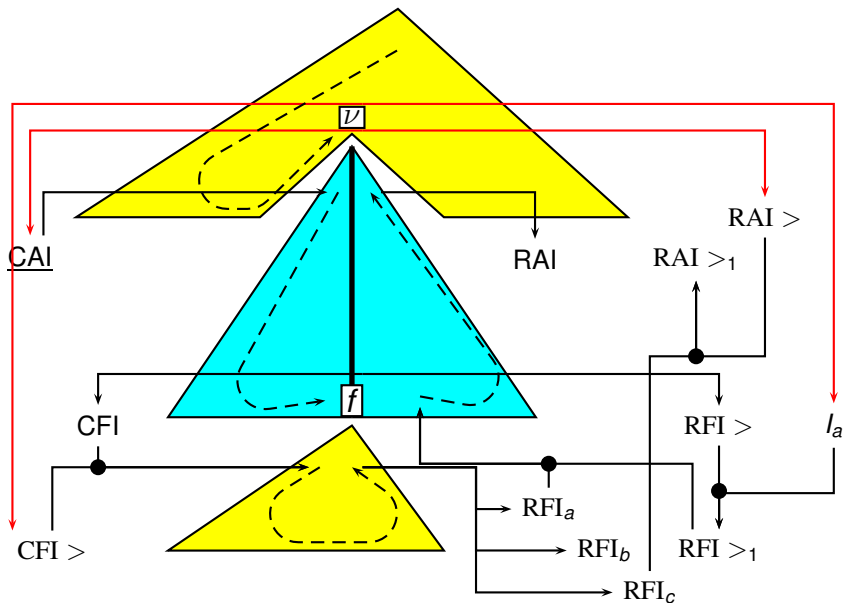
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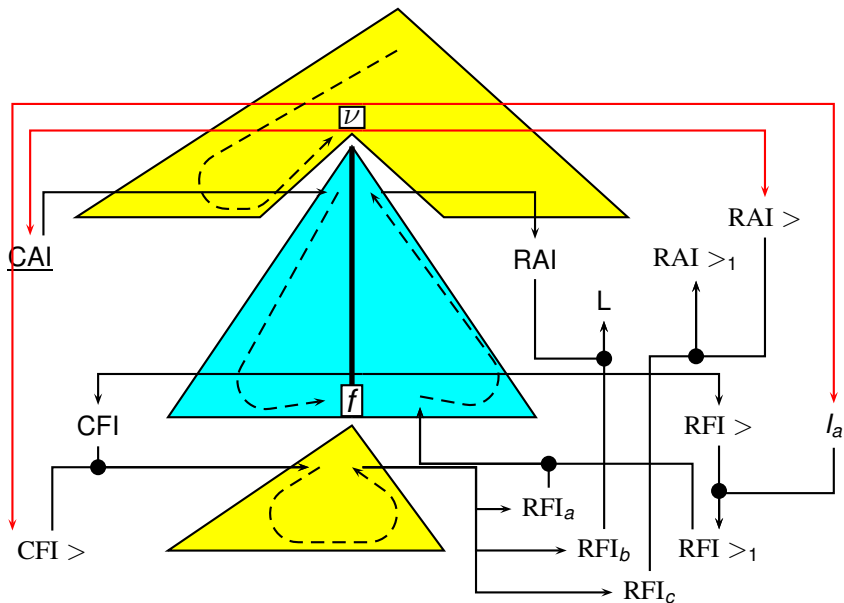
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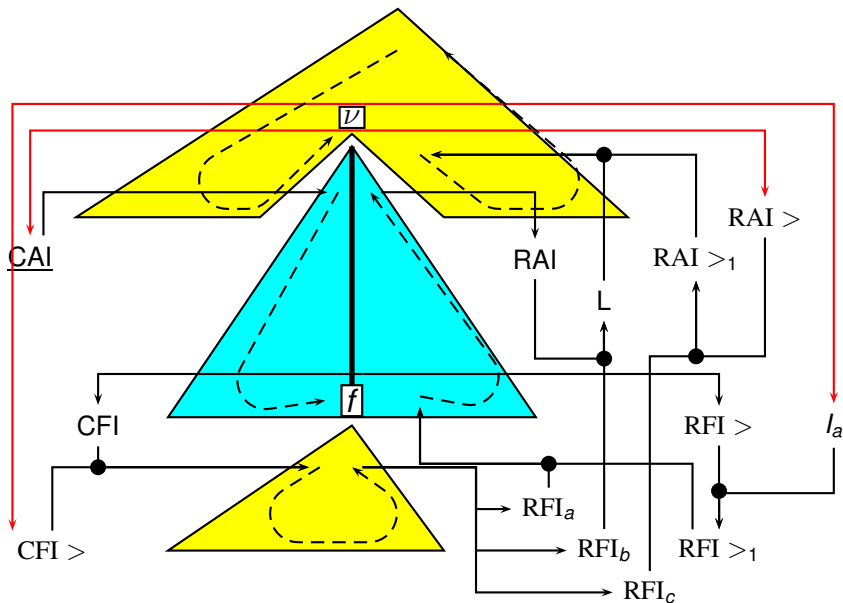


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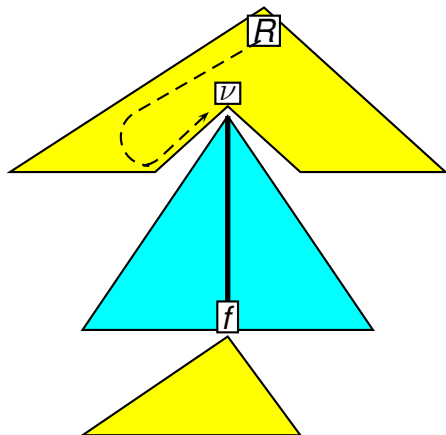




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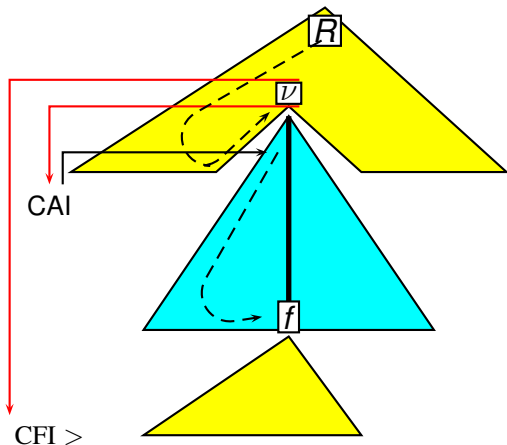


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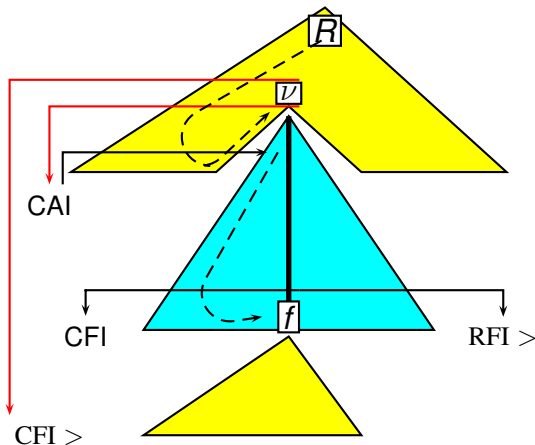
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- But more efficient in practice !
- And suggesting extensions, based on the idea of continuation

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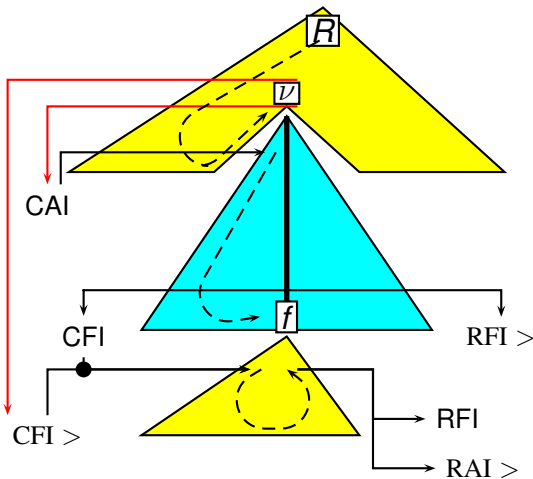
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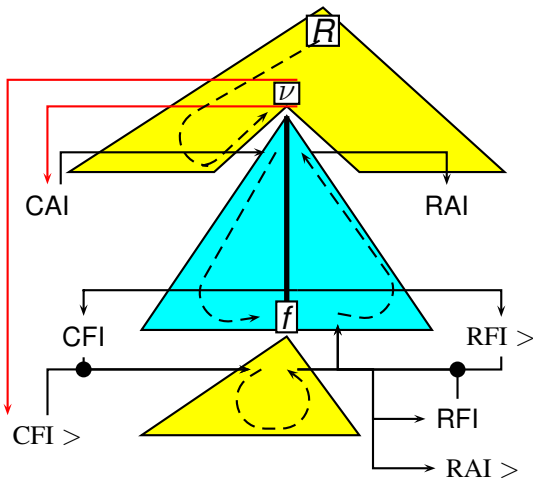
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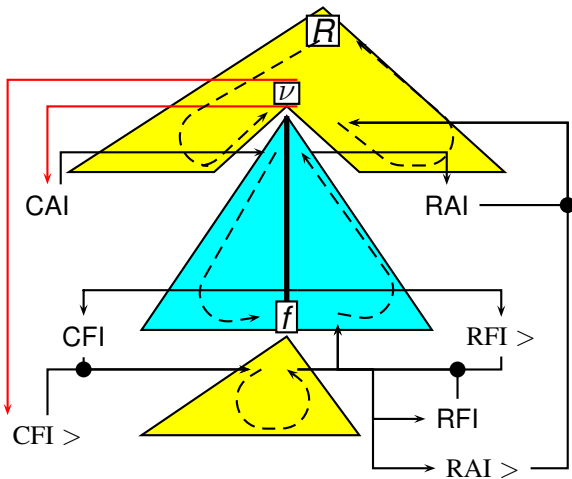
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# Part II

## MCS in general



- 4 Thread Automata and MCS formalisms
- 5 A Dynamic Programming interpretation for TAs

# Mildly Context Sensitivity

An informal notion covering formalisms such that:

- they are powerful enough to model **crossing**, such as  $a^n b^n c^n$
- they are parsable with **polynomial complexity**  
i.e. Given  $L$ , there exists  $k$ , membership  $w \in L$  checked in  $O(|w|^k)$
- they generate string languages satisfying the **constant growth property**  
$$\exists G, G \text{ finite}, \exists n_0, \forall w \in \mathcal{L}, |w| > n_0 \implies \exists g \in G, \exists w' \in \mathcal{L}, |w| = |w'| + g$$
  
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Some MCS languages:

- TAGs and LIGs
- Local Multi Component TAGs (MC-TAGs **Weir**)
- Linear Context-Free Rewriting Systems (LCFRS **Weir**)
- Simple Range Concatenation Grammars (sRCG **Boullier**)

The Constant Growth property subsumed by stronger **semi-linearity** under Parikh image

The Parikh image of  $w \in \{a_1, \dots, a_n\}^*$  defined as  $p(w) = (|w|_{a_1}, \dots, |w|_{a_n})$

The Parikh image of  $L$  defined as  $p(L) = \{p(w) \mid w \in L\}$

A set  $V$  of vectors over  $\mathbb{N}^\Sigma$  is linear is generated by a base  $v_0, v_1, \dots, v_n \in \mathbb{N}^\Sigma$  by

$$V = \{v_0 + \sum_{i=1}^n k_i v_i \mid k_i \in \mathbb{N}\}$$

$V$  is semilinear if  $V = \bigcup_{i=1}^k V_i$  is a finite union of linear sets  $V_i$

A language  $L$  is semilinear if  $p(L)$  is semilinear

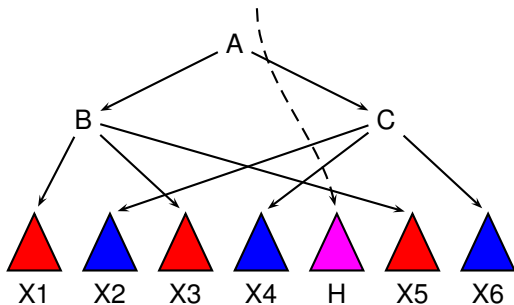
(intuition) A MCS language is generated, modulo some **permutations**, by a finite set of generators

# MCS: discontinuity and interleaving

Discontinuous interleaved constituents present in linguistic phenomena  
Nesting, Crossing, Topicalization, Deep extraction, Complex Word-Order . . .

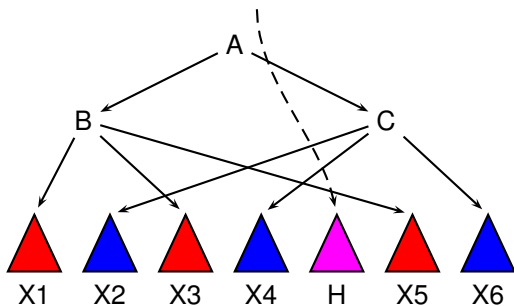
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- **LFCRS:**  $A \leftarrow f(B, C)$ ,  $f$  linear non erasing function on string tuples.

$$f(\langle x_1, x_3, x_5 \rangle, \langle x_2, x_4, x_6 \rangle) = \langle x_1 x_2 x_3 x_4, x_5 x_6 \rangle$$

- **sRCG**  $A(x_1 \cdot x_2 \cdot x_3 \cdot x_4, x_5 \cdot x_6) \leftarrow B(x_1, x_3, x_5), C(x_2, x_4, x_6)$   
range variables  $x_i$ ; concatenation “.”; holes “,”

*Linear Context-Free Rewriting Systems (LCFRS)* , a restricted form of *generalized CFGs*

A LCFRS is a tuple  $G = (\mathcal{N}, \Sigma, S, \mathcal{P}, \mathcal{F})$  where

- $\mathcal{P}$  is a finite set of productions as follows, with  $f \in \mathcal{F}$

$$A \leftarrow f(A_1, \dots, A_n)$$

- $\mathcal{F}$  is a set of linear regular operations over tuples of strings in  $\Sigma^*$

$$f(\langle x_{1,1}, \dots, x_{1,k_1} \rangle, \dots, \langle x_{n,1}, \dots, x_{n,k_n} \rangle) = \langle t_1, \dots, t_k \rangle$$

where  $V = \{x_{i,j}\}$  are variables (over  $\Sigma^*$ ) and  $t_i \in (\Sigma \cup V)^*$  and

- ▶ (regular or non-erasing)  $\forall x_{i,j}, \exists t_u, x_{i,j} \in t_u$
- ▶ (linear)  $\forall x_{i,j}, x_{i,j} \in t_u \wedge x_{i,j} \in t_v \implies u = v$

Assuming  $\text{arity}(S) = 1$ ,

$$L(G) = \{w \mid S \implies \langle w \rangle\}$$

where

$$\begin{aligned} A &\implies f() && \text{if } A \rightarrow f() \in \mathcal{P} \\ A &\implies f(t_1, \dots, t_n) && \text{if } A \rightarrow f(A_1, \dots, A_n) \in \mathcal{P} \wedge \forall i, A_i \implies t_i \end{aligned}$$



Range Concatenation Grammars (RCG) [Boullier] :

Constraints on intervals on the input string.

For language  $a^n b^n c^n$

$S(X @ Y @ Z) \rightarrow A(X, Y, Z) .$   
 $A("a" @ X, "b" @ Y, "c" @ Z) \rightarrow$   
 $A(X, Y, Z) .$   
 $A(" ", " ", " ") \rightarrow .$

$aabbcc \quad S([0, 6]) \rightarrow$   
 $aabbcc \quad A([0, 2], [2, 4], [4, 6]) \rightarrow$   
 $aabbcc \quad A([1, 2], [3, 4], [5, 6]) \rightarrow$   
 $aabbcc \quad A([2, 2], [4, 4], [6, 6]) \rightarrow$

RCG is an operational formalism for encoding linguistic formalisms where discontinuous constituents are used.

RCG allow modular grammar writing

**concatenation**  $G(X @ Y) \rightarrow G_1(X), G_2(Y).$

**union**  $G(X) \rightarrow G_1(X) \mid G_2(X).$

**intersection**  $G(X) \rightarrow G_1(X), G_2(X).$

Linear non-erasing positive RCGs equivalent to LCFRS

Full RCGs are PTIME (equivalent to Datalog)

- MCS have theoretical polynomial complexity  $O(n^u)$  depending upon
  - ▶ degree of **discontinuity**, (also **fanout**, **arity**)
  - ▶ degree of **interleaving**, (also **rank**)
- But no uniform framework to express parsing strategies and tabular algorithms
  - ▶ operational device: **Deterministic Tree Walking Transducer** (**Weir**), but no tabular algorithm
  - ▶ operational formalism **sRCG** with tabular algorithm (**Boullier**) but not for **prefix-valid strategies**

Notion of **Thread Automata** to model discontinuity and interleaving through the suspension/resume of threads.

TWA may be used to check properties of (binary) trees  
(by accepting or rejecting them)

A (non-deterministic) TWA is a tuple  $A = (\mathcal{Q}, \Sigma, I, F, R, \delta)$  where

- $\mathcal{Q}$  is a finite set of states
- $\Sigma$  a finite set of node labels
- $I, F, R \subset \mathcal{Q}$  the initial, accepting, rejecting states
- $\delta$  the finite set of transitions in  $\mathcal{Q} \times \Sigma \times \text{Pred} \times \text{Dirs} \times \mathcal{Q}$  where
  - ▶  $\text{Pred} \subset \{\text{root}, \text{left}, \text{right}, \text{leaf}\}$  is a set of predicates for testing nodes
  - ▶  $\text{Dirs} \subset \{\text{stay}, \text{up}, \text{left}, \text{right}\}$  a set of directions

Deterministic TWA:  $\delta : \mathcal{Q} \times \Sigma \times \text{Pred} \mapsto \text{Dirs} \times \mathcal{Q}$

Given a  $\Sigma$ -tree  $\tau = (V, E)$ , a configuration is given by  $(\nu, q) \in V \times \mathcal{Q}$

**Extensions:** Pebble Automata ([Engelfriet](#))

# Tree Walking Transducers

Similar to TWA, but emits strings when walking over a tree (in some tree set)

A deterministic TWD (**Weir**) is a tuple  $T = (\mathcal{Q}, G, \Sigma_O, q_I, F, \delta)$  where

- $G = (\mathcal{N}, \Sigma_I, \mathcal{S}, \mathcal{P})$  is a CFG
- $\Sigma_O$  a finite set of output symbols
- $\delta : \mathcal{Q} \times (\mathcal{N} \cup \Sigma_I \cup \{\epsilon\}) \mapsto \text{Dirs} \times \mathcal{Q} \times \Sigma_O$   
with  $\text{Dirs} = \{\text{stay}, \text{up}, \text{down}_1, \dots, \text{down}_n\}$

A transition step given by

$$(q, \gamma, \nu, w) \xrightarrow{\star} (q', \gamma, \nu', w.v) \text{ if } \begin{cases} (q, \text{dir}) = \delta(q, \text{label}(\nu)) \\ \nu' = \text{dir}(\nu) \end{cases}$$

The language generated by  $T$  defined as

$$L(T) = \{w \mid (q_I, \gamma, r_\gamma, \epsilon) \xrightarrow{\star} (q_f, \gamma, \uparrow, w)\}$$

with  $q_f \in F$ ,  $\gamma$  a derivation tree for  $G$  with root  $r_\gamma$

and  $\uparrow$  a virtual node parent of  $r_\gamma$

Weir's result:  $L(\text{DTWD}) = \text{LCFRL}$

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- create a subthread  $p.u$  for a sub-constituent [PUSH]
- suspend thread at constituent discontinuity, and (**resume**) either the parent thread [SPOP] or some direct subthread [SPUSH]
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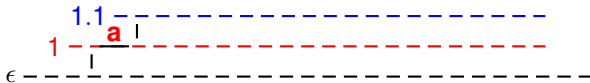


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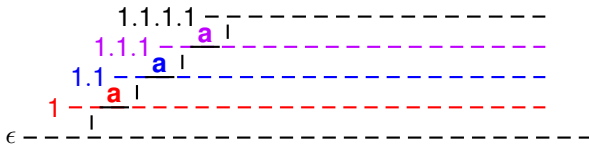


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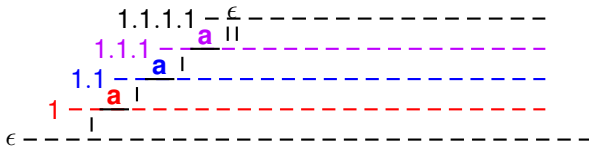


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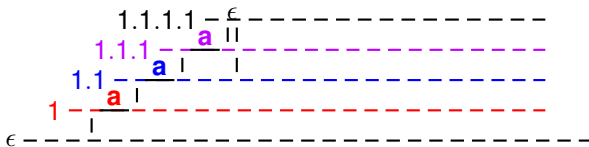


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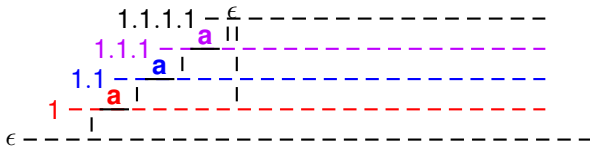


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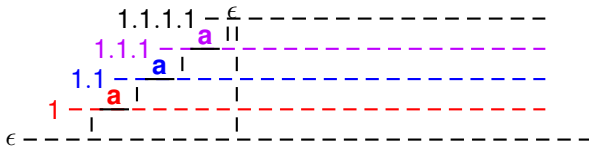


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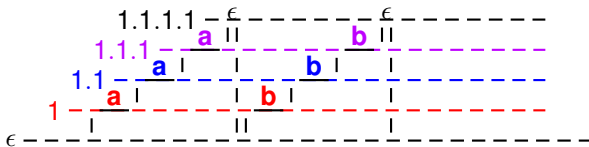


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**Configuration**  $\langle \text{position } l, \text{ active thread path } p, \text{ thread store } \mathcal{S} = \{p_i:A_i\} \rangle$   
 $\mathcal{S}$  closed by prefix:  $p.u \in \text{dom}(\mathcal{S}) \implies p \in \text{dom}(\mathcal{S})$   
Note: **stateless** automata (but no problem for variants with states)

**Triggering function**  $a = \Phi(A)$  amount of information needed to trigger transitions.  
 $\implies$  useful to get linear complexity  $O(|G|)$  w.r.t. grammar size  $|G|$   
Default:  $\Phi = \text{Identity}$

**Driver function**  $u \in \delta(A)$  Drive thread creations and suspensions  
 $\implies$  reduce number of transitions  
(TA variants without  $\delta$  should be possible)



# Formal presentation of TA (cont'd)

**SWAP**  $B \xrightarrow{\alpha} C$  : Changes the content of the active thread, possibly scanning a terminal.

$$\langle l, p, S \cup p:B \rangle \xrightarrow{\tau} \langle l + |\alpha|, p, S \cup p:C \rangle \quad a_l = \alpha \text{ if } \alpha \neq \epsilon$$

**PUSH**  $b \mapsto [b]C$  : Creates a new subthread (unless present)

$$\langle l, p, S \cup p:B \rangle \xrightarrow{\tau} \langle l, pu, S \cup p:B \cup pu:C \rangle \quad (b, u) \in \Phi\delta(B) \wedge pu \notin \text{dom}(S)$$

**POP**  $[B]C \mapsto D$  : Terminates thread  $pu$  (if no existing subthreads).

$$\langle l, pu, S \cup p:B \cup pu:C \rangle \xrightarrow{\tau} \langle l, p, S \cup p:D \rangle \quad pu \notin \text{dom}(S)$$

**SPUSH**  $b[C] \mapsto [b]D$  : Resumes the subthread  $pu$  (if already created)

$$\langle l, p, S \cup p:B \cup pu:C \rangle \xrightarrow{\tau} \langle l, pu, S \cup p:B \cup pu:D \rangle \quad (b, u^s) \in \Phi\delta(B)$$

**SPOP**  $[B]c \mapsto D[c]$  : Resumes the parent thread  $p$  of  $pu$

$$\langle l, pu, S \cup p:B \cup pu:C \rangle \xrightarrow{\tau} \langle l, p, S \cup p:D \cup pu:C \rangle \quad (c, \perp) \in \Phi\delta(C)$$

# Characterizing Thread Automata

Key parameters:

$h$  maximal number of suspensions to the parent thread  
 $h$  finite ensures termination (of tabular parsing)

$d$  maximal number of simultaneously *alive* subthreads

$l$  maximal number of subthreads

$s$  maximal number of suspensions (parent + alive subthreads)

$$s \leq h + dh \leq h + lh$$

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Worst-case Complexity:

$$\left. \begin{array}{l} \text{space } O(n^u) \\ \text{time } O(n^{1+u}) \end{array} \right\} \text{ where } \left\{ \begin{array}{l} u = 2 + s + x \\ x = \min(s, (l - d)(h + 1)) \end{array} \right.$$

$$\Rightarrow \left\{ \begin{array}{l} \text{space between } O(n^{2+2s}) \text{ and [when } l = d] O(n^{2+s}) \\ \text{time between } O(n^{3+2s}) \text{ and [when } l = d] O(n^{3+s}) \end{array} \right.$$

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 $h$  finite ensures termination (of tabular parsing)

$d$  maximal number of simultaneously *alive* subthreads

$l$  maximal number of subthreads

$s$  maximal number of suspensions (parent + alive subthreads)

$$s \leq h + dh \leq h + lh$$

Worst-case Complexity:

$$\left. \begin{array}{l} \text{space } O(n^u) \\ \text{time } O(n^{1+u}) \end{array} \right\} \text{ where } \left\{ \begin{array}{l} u = 2 + s + x \\ x = \min(s, (l - d)(h + 1)) \end{array} \right.$$

$$\Rightarrow \left\{ \begin{array}{l} \text{space between } O(n^{2+2s}) \text{ and [when } l = d] O(n^{2+s}) \\ \text{time between } O(n^{3+2s}) \text{ and [when } l = d] O(n^{3+s}) \end{array} \right.$$

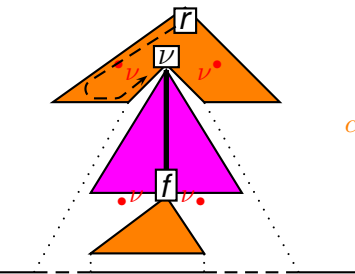
Push-Down Automata (PDA) for CFG  $\equiv$  TA( $h=0, d=1, s=0$ )

$$\Rightarrow \text{space } O(n^2) \text{ and time } O(n^3)$$

**Idea:** Assign a thread per elementary tree **traversal** (substitution or adjunction)  
Suspend and return to parent thread to handle a foot node

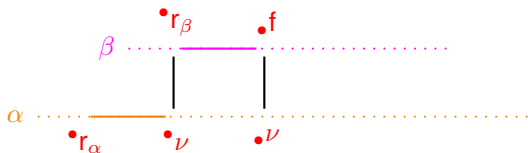
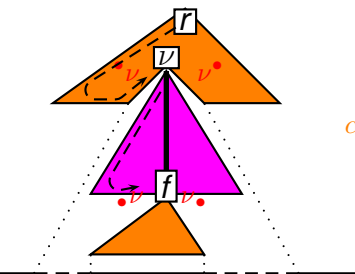
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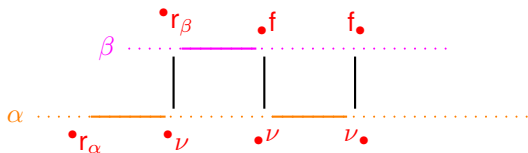
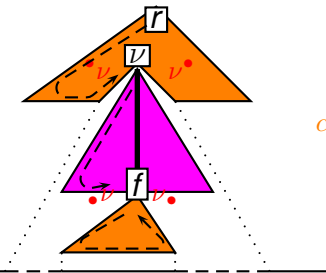
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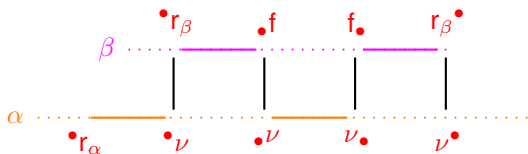
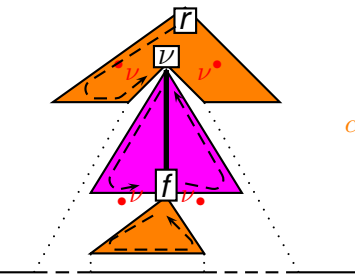
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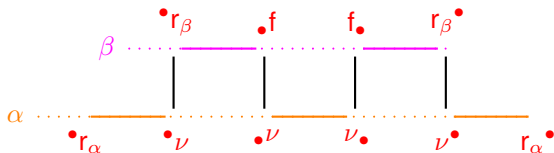
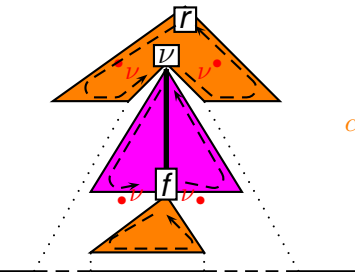
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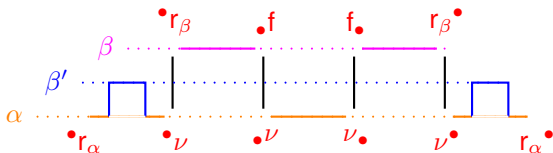
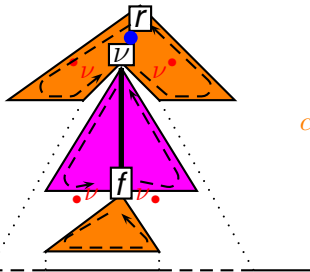
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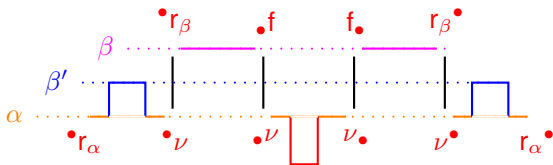
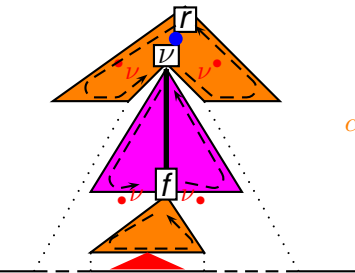
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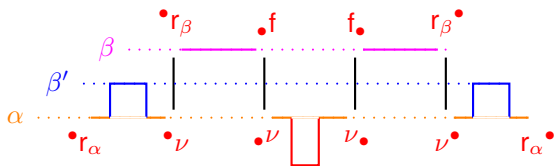
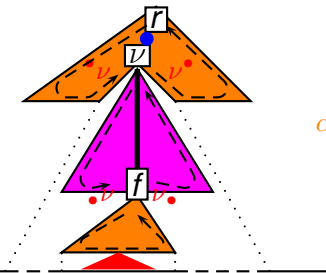
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One thread per tree  $h = 1, d = \max(\text{depth}(\text{trees}))$   
 $\Rightarrow [s = 1 + d]$  space  $O(n^{4+2d})$  and time  $O(n^{5+2d})$

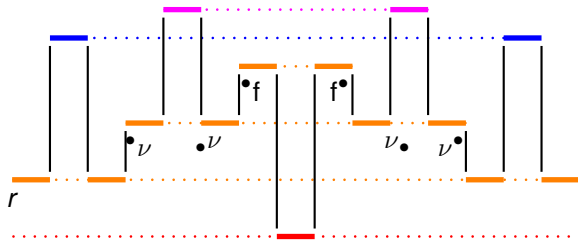
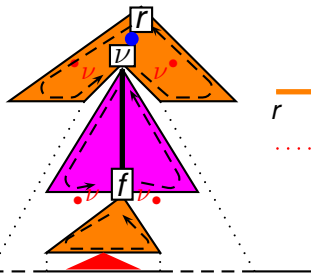
# Parsing TAG: an alternate parsing strategy

Using more than one thread per elementary tree: 1 thread per subtree ( $\sim$  LIG)

$\Rightarrow$  implicit extraction of subtrees

$\Rightarrow$  implicit normal form (using a third kind of tree operation)

$\Rightarrow$  usual  $n^6$  time complexity



**Note:** Similar to a TAG encoding in RCG proposed by [Boullier](#)

Always possible to reduce the number of live subthreads (down to 2).

- if a thread  $p$  has  $d + 1$  subthreads, add a new subthread  $p.v$  that inherits  $d$  subthreads of  $p$
- generally increases the number of parent suspensions  $h$
- but may also exploit good topological properties, such as [well-nesting](#) (TAGs).

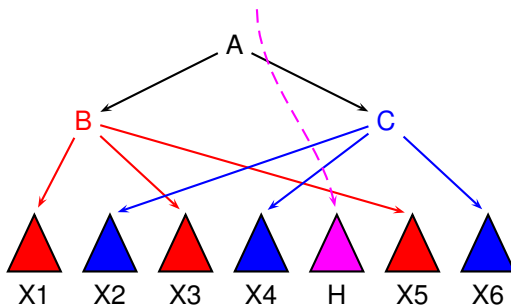


# Parsing (ordered simple) RCG

Range Concatenation Grammars (Boullier)

$\gamma : A(X_1 X_2 X_3 X_4, X_5 X_6) \longrightarrow B(X_1, X_3, X_5)C(X_2, X_4, X_6)$

Ordered simple RCGs  $\equiv$  Linear Context-Free Rewriting Systems (LCFRS)

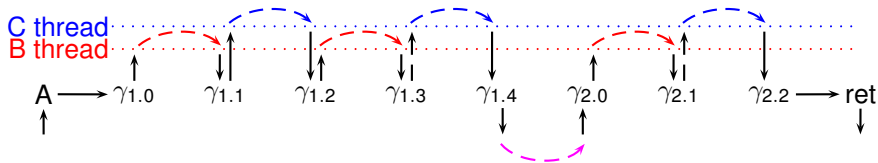
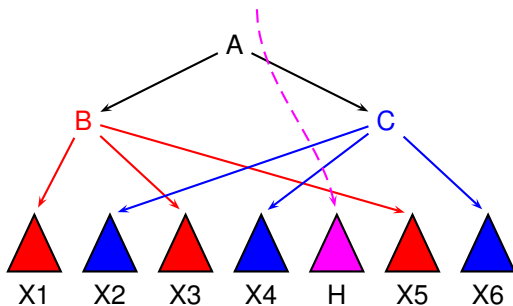


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**Idea:** assign a thread to traverse (in any order) the elementary trees of a set  $\Sigma$ , using extended dotted nodes  $\Sigma:\rho\sigma$  where

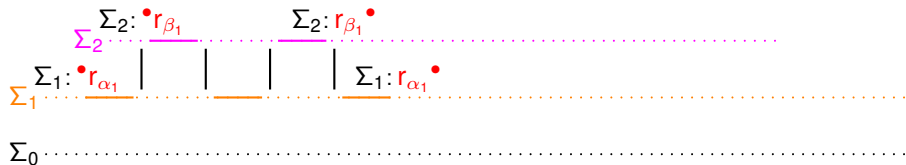
- $\rho$  stack of dotted nodes of trees being traversed
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# Parsing (set-local) MC-TAGs

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Eg.: Adjoin trees of set  $\Sigma_2 = \{\beta_1, \beta_2\}$  on nodes of trees of set  $\Sigma_1 = \{\alpha_1, \alpha_2\}$

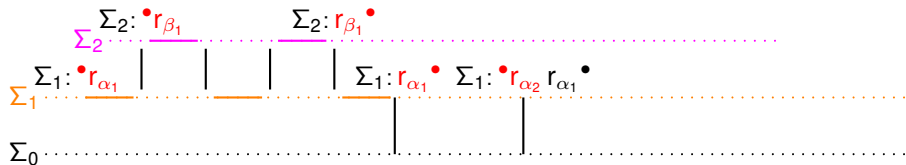


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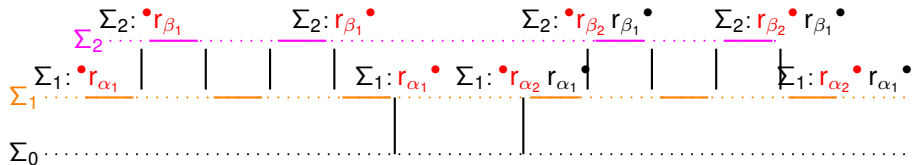


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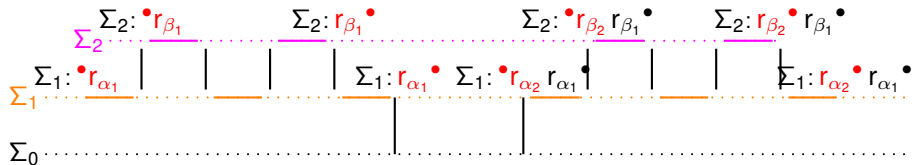


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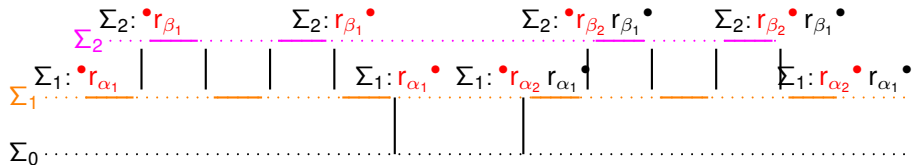


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Time complexity  $O(n^{3+2(m+v)})$  where  $\begin{cases} m \text{ max number of trees per set} \\ v \text{ max number of nodes per set} \end{cases}$



4 Thread Automata and MCS formalisms

5 A Dynamic Programming interpretation for TAs

Direct evaluation of TA  $\rightsquigarrow$  exponential complexity and non-termination

Use tabular techniques based on [Dynamic Programming](#) interpretation of TAs:

**Principle:** Identification of a class of subderivations that

- may be tabulated as compact [items](#), removing non-pertinent information
- may be combined together and with transitions to retrieve all derivations

Methodology followed for PDAs (CFGs) and 2SAs (TAGs)

DP interpretation of TA derivations:

(Tabulated) Item  $\equiv$  pertinent information about an (active) thread

1– Start point

3– (current) Parent suspensions

2– (current) End point

4– (current) Subthread suspensions for **live** subthreads

# Dynamic Programming – Items

DP interpretation of TA derivations:

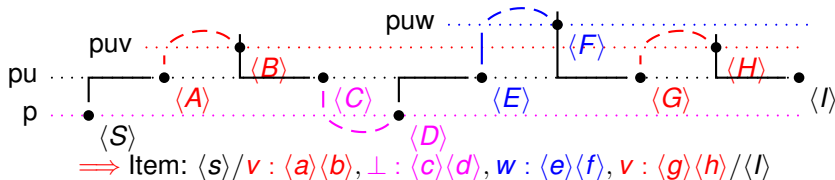
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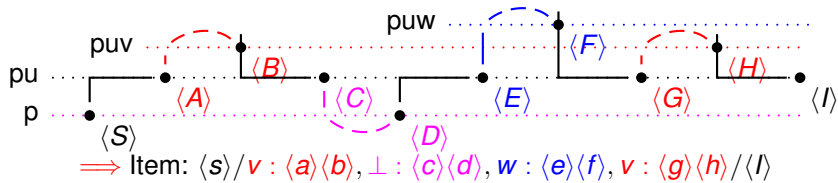
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Projection  $x = \Phi(X)$  used to trigger transition applications

$\Rightarrow$  easy way to get complexity  $O(|G|)$

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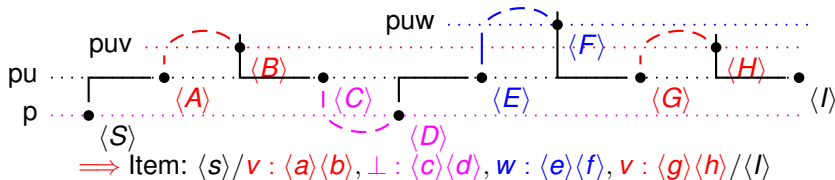
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Projection  $x = \Phi(X)$  used to trigger transition applications

$\Rightarrow$  easy way to get complexity  $O(|G|)$

Space complexity:

- at most 2 indices per suspensions + start + end =  $2(1 + s) \leq 2(1 + h + dh)$
- Scanning parts generally of fixed length (independent of  $n$ )

$\Rightarrow$  1 index per suspension

# Dynamic Programming – Application rules

Based on following model:

parent item   **son item**   **trans**

parent or son extension

{fitting son and parent items}

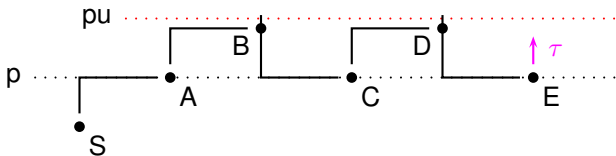
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Case [SPUSH]: parent item **down-extends** son item





# Dynamic Programming – Application rules

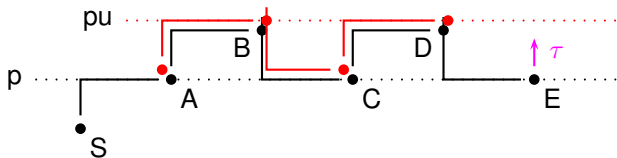
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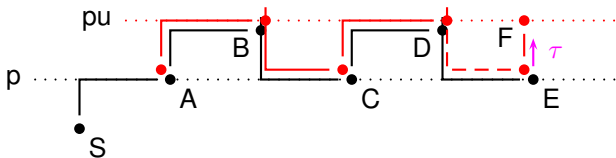
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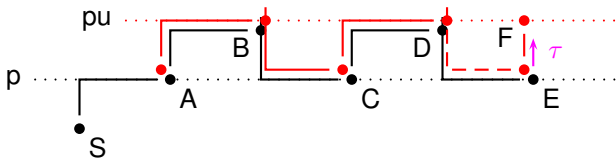
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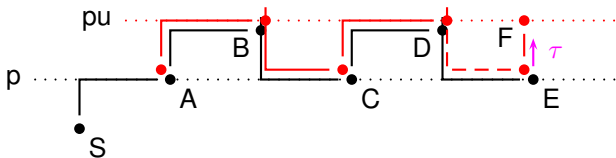
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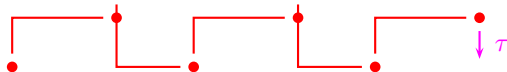
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Case [SPUSH]: parent item **down-extends** son item



Case [SPOP]: son item **up-extends** parent item





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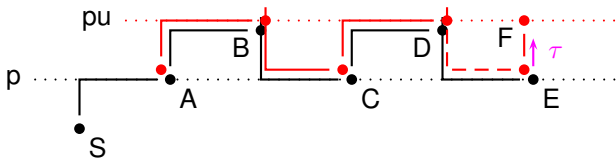
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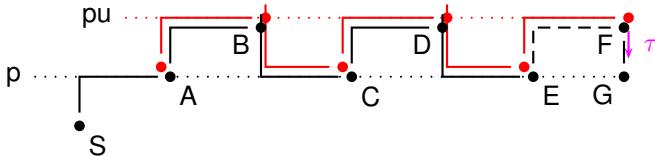
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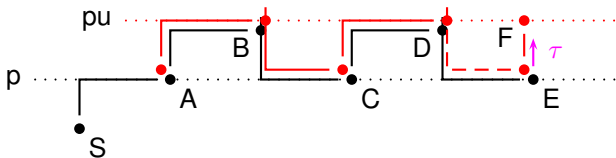
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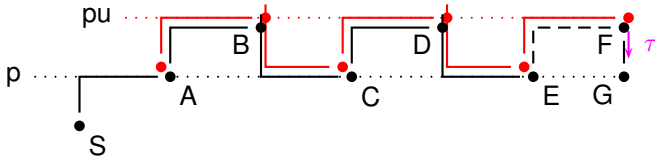
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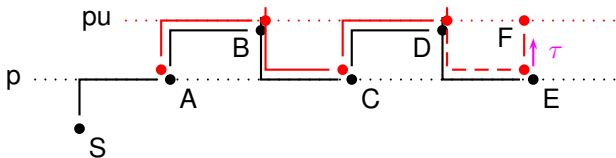
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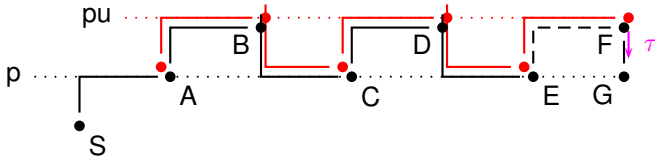
parent or son extension

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Case [SPUSH]: parent item down-extends son item



Case [SPOP]: son item up-extends parent item



Time complexity: all indices of parent item + end position of son item  
ignore indices of son item not related to parent suspensions



$$\frac{B \xrightarrow{\alpha} C \quad \langle a \rangle / S / \langle B \rangle}{\langle a \rangle / S / \langle C \rangle} \quad a_r = \alpha \text{ if } \alpha \neq \epsilon \quad (\text{SWAP})$$

$$\frac{b \mapsto [b]C \quad \star / \star // \langle B \rangle^I}{\langle b \rangle // \langle C \rangle} \quad \{ (b, u) \in \Phi\delta(B) \wedge u \notin \text{ind}(I) \} \quad (\text{PUSH})$$

$$\frac{[B]C \mapsto D \quad \langle a \rangle / S / \langle B \rangle^I \quad J}{\langle a \rangle / S_{/u} / \langle D \rangle} \quad \left\{ \begin{array}{l} J \xrightarrow{u} I \wedge (b, u) \in \Phi\delta(B) \\ J^\bullet = \langle C \rangle \wedge \text{ind}(J) \subset \{\perp\} \end{array} \right. \quad (\text{POP})$$

$$\frac{b[C] \mapsto [b]D \quad I \quad \langle a \rangle / S / \langle C \rangle^J}{\langle a \rangle / S, \perp : \langle c \rangle \langle b \rangle // \langle D \rangle} \quad \left\{ \begin{array}{l} I \searrow_u J \wedge I^\bullet = \langle B \rangle \\ (b, u) \in \Phi\delta(B) \wedge (c, \perp) \in \Phi\delta(C) \end{array} \right. \quad (\text{SPUSH})$$

$$\frac{[B]c \mapsto D[c] \quad \langle a \rangle / S / \langle B \rangle^I \quad J}{\langle a \rangle / S, u : \langle b \rangle \langle c \rangle // \langle D \rangle} \quad \left\{ \begin{array}{l} J \xrightarrow{u} I \wedge (b, u) \in \Phi\delta(B) \\ J^\bullet = \langle C \rangle \wedge (c, \perp) \in \Phi\delta(C) \end{array} \right. \quad (\text{SPOP})$$