

Language and Statistics II

Lecture 13: Deductive Parsing,
Especially with Weights

Noah Smith

Remember Prolog?

Invented around 1972 for AI and CL.

Horn clauses:

```
father(X,Y) :- parents(X,_,Y), male(X) .
```

```
grandfather(X,Z) :-  
    father(X,Y), father(Y,Z) .
```

```
male(john) .
```

```
male(joe) .
```

```
parents(william,jane,john) .
```

```
parents(joe,mary,william) .
```

Prolog

- Given a query, Prolog tries to prove some instantiation of it.

```
?- grandpa (X, john) .
```

```
X=joe;
```

```
yes.
```

- Prolog uses a particular brand of **search**, which we won't talk about.
- The point: deduction and theorem proving are a useful way to **describe** algorithms in CL.

Parsing as Deduction (high-level view)

- *Axioms*: the sentence s , the grammar G
- *Inference rules*: correspond to the parsing algorithm
- *Theorems*: partial parses, then stored in **chart**
- *Stopping criterion*: “goal” reached or agenda empty
- *Output*: true iff $s \in L(G)$ (i.e., “goal” is proven), along with a proof

Shieber, Schabes, & Pereira (1995): use Prolog!

- For efficiency, minimize redundancy in chart and agenda.

Example 1: top-down parsing

Item form: $[\bullet\beta, j]$

Axioms: $[\bullet S, 0]$

Goals: $[\bullet, n]$

Inference rules:

Scanning
$$\frac{[\bullet w_{j+1}\beta, j]}{[\bullet\beta, j+1]}$$

Prediction
$$\frac{[\bullet B\beta, j]}{[\bullet\gamma\beta, j]} \quad B \rightarrow \gamma$$

$[\bullet\beta, j]$ means: “Starting after word j , I need to build β (a sequence of symbols).”

Scanning: match the next word in the sentence to the next required symbol.

Prediction: use grammar to see how next required symbol could expand.

What would happen if we implemented this?

Example 2: shift-reduce parsing

Item form: $[\alpha \bullet, j]$

Axioms: $[\bullet, 0]$

Goals: $[S \bullet, n]$

Inference Rules:

Shift
$$\frac{[\alpha \bullet, j]}{[\alpha w_{j+1} \bullet, j+1]}$$

Reduce
$$\frac{[\alpha \gamma \bullet, j]}{[\alpha B \bullet, j]} \quad B \rightarrow \gamma$$

$[\alpha \bullet, j]$ means: “Up to word j , I have begun to build α (a stack of symbols).”

Shift (scanning): move the next word in the sentence to the stack.

Reduce (completion): use grammar to see how the top symbols on the stack can be combined.

Algorithm	<i>Bottom-Up</i>	<i>Top-Down</i>	<i>Earley's</i>
Item form	$[\alpha \bullet, j]$	$[\bullet \beta, j]$	$[i, A \rightarrow \alpha \bullet \beta, j]$
Invariant	$\alpha w_{j+1} \cdots w_n \xrightarrow{*} w_1 \cdots w_n$	$S \xrightarrow{*} w_1 \cdots w_j \beta$	$S \xrightarrow{*} w_1 \cdots w_i A \gamma$ $\alpha w_{j+1} \cdots w_n \xrightarrow{*} w_{i+1} \cdots w_n$
Axioms	$[\bullet, 0]$	$[\bullet S, 0]$	$[0, S' \rightarrow \bullet S, 0]$
Goals	$[S \bullet, n]$	$[\bullet, n]$	$[0, S' \rightarrow S \bullet, n]$
Scanning	$\frac{[\alpha \bullet, j]}{[\alpha w_{j+1} \bullet, j+1]}$	$\frac{[\bullet w_{j+1} \beta, j]}{[\bullet \beta, j+1]}$	$\frac{[i, A \rightarrow \alpha \bullet w_{j+1} \beta, j]}{[i, A \rightarrow \alpha w_{j+1} \bullet \beta, j+1]}$
Prediction		$\frac{[\bullet B \beta, j]}{[\bullet \gamma \beta, j]} \quad B \rightarrow \gamma$	$\frac{[i, A \rightarrow \alpha \bullet B \beta, j]}{[j, B \rightarrow \bullet \gamma, j]} \quad B \rightarrow \gamma$
Completion	$\frac{[\alpha \gamma \bullet, j]}{[\alpha B \bullet, j]} \quad B \rightarrow \gamma$		$\frac{[i, A \rightarrow \alpha \bullet B \beta, k] \quad [k, B \rightarrow \gamma \bullet, j]}{[i, A \rightarrow \alpha B \bullet \beta, j]}$

From Shieber, Schabes, & Pereira (1995)

Earley's Algorithm, etc.

- A more efficient combination of two naïve algorithms. (A more efficient logic program.)
- SS&P went on to give logic programs for CCG parsing and TAG parsing.
- McAllester (2002) shows how to automatically derive asymptotic runtime and space bounds for such programs.

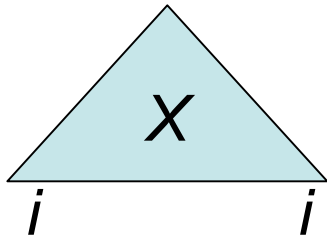
The Point

- Logic programs are helpful for
 - Algorithm specification
 - Prototype implementation
 - Finding **more efficient** algorithms

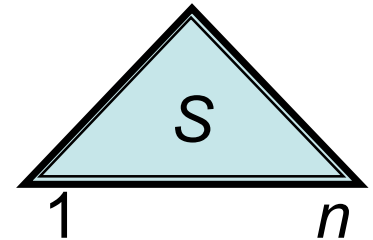
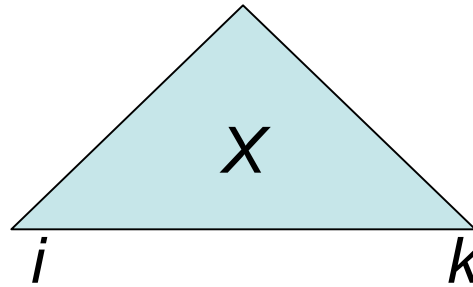
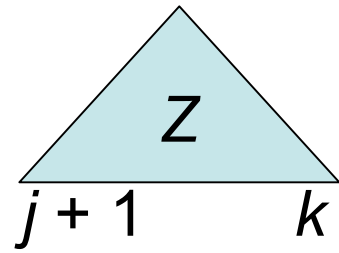
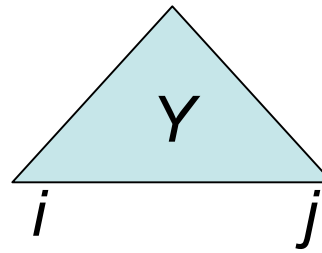
CKY

$X \rightarrow w_i$

w_i

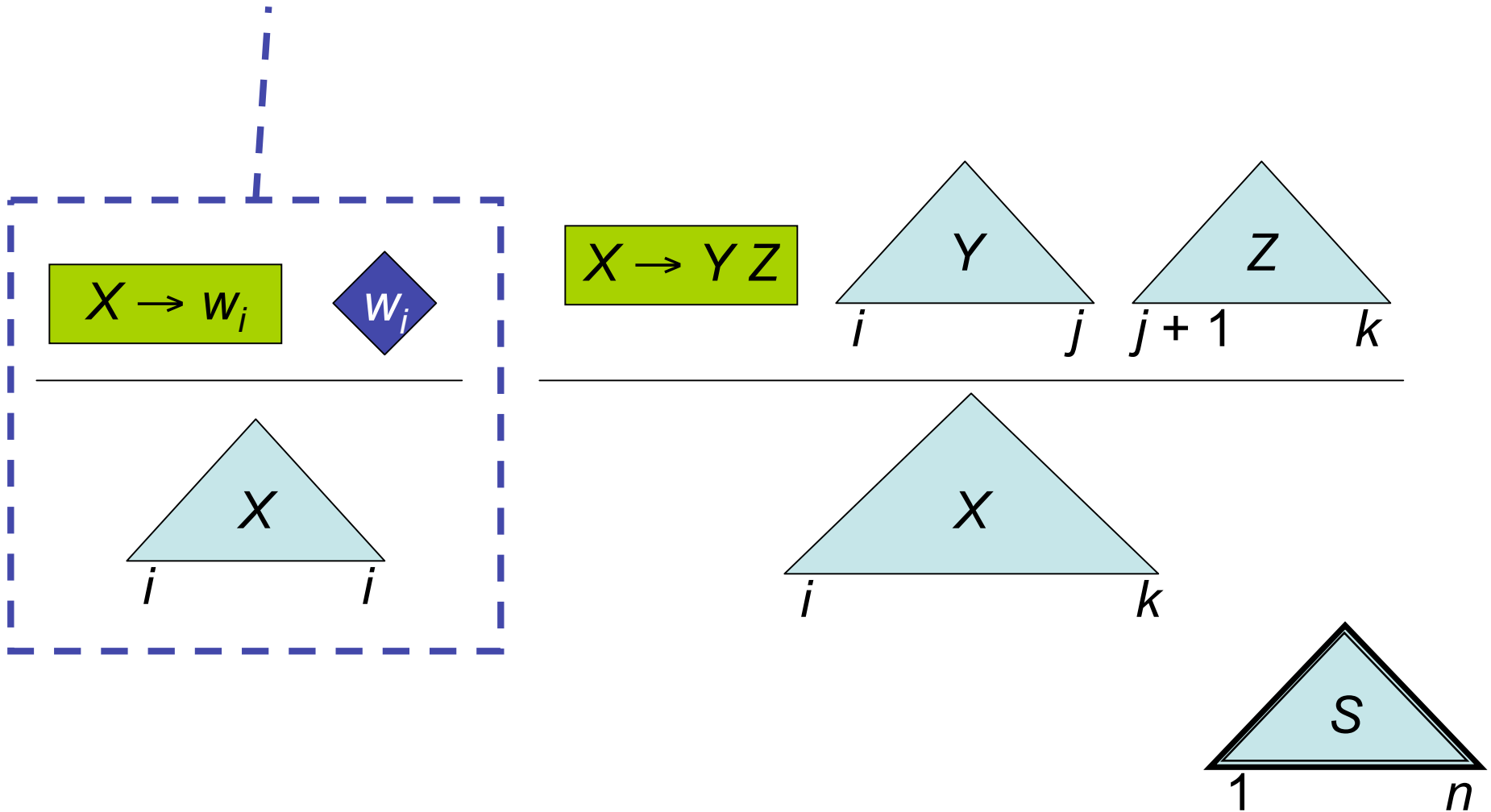


$X \rightarrow YZ$



CKY

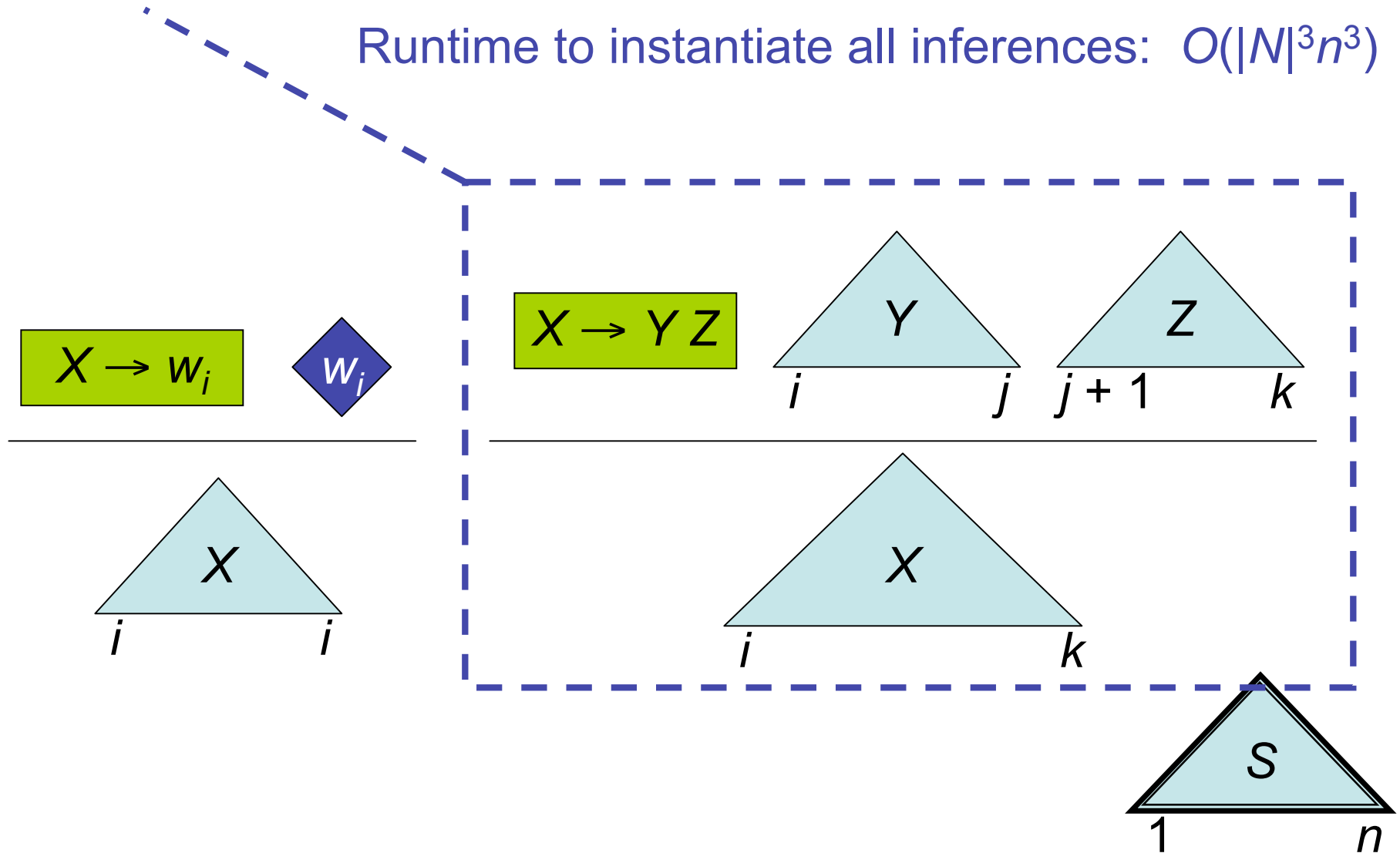
$\{X, i\}$ Runtime to instantiate all inferences: $O(|N|n)$



CKY

$\{X, Y, Z, i, j, k\}$

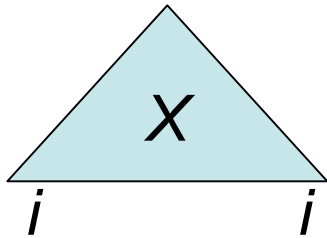
Runtime to instantiate all inferences: $O(|N|^3 n^3)$



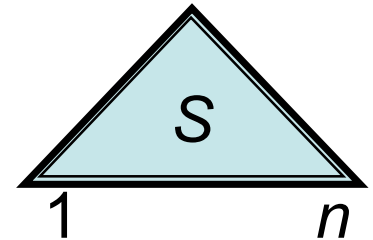
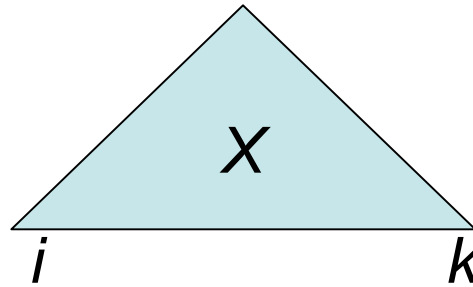
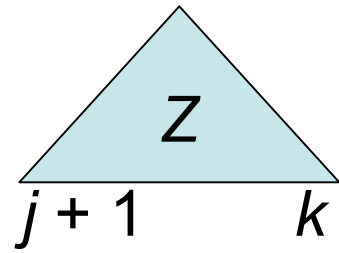
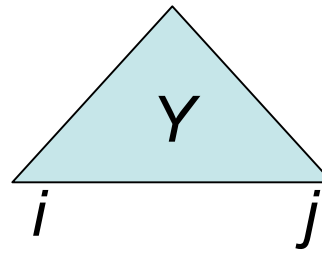
CKY

$X \rightarrow w_i$

w_i

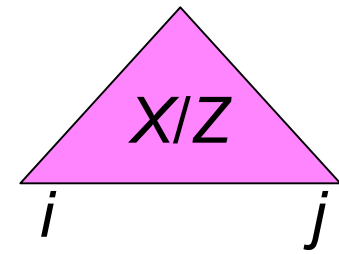
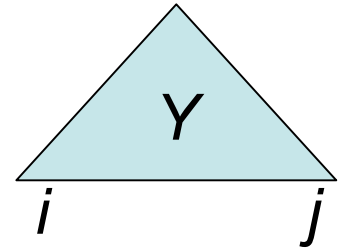


$X \rightarrow YZ$

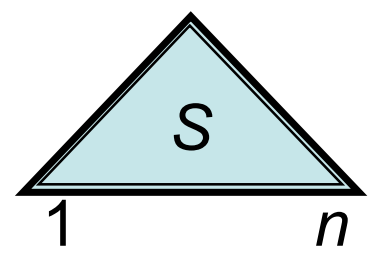
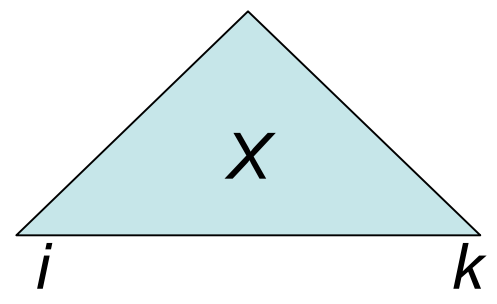
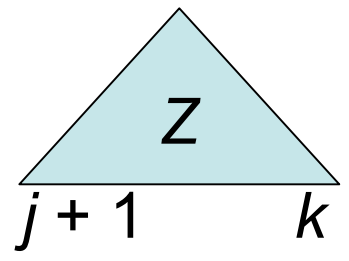
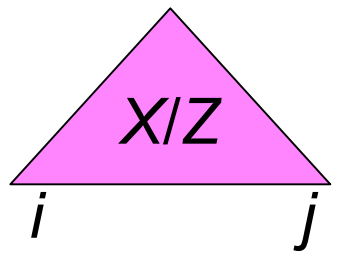
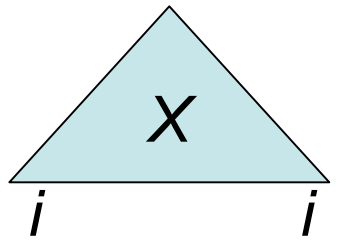


CKY

$X \rightarrow YZ$

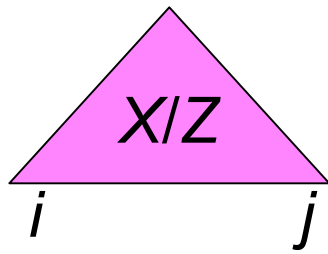
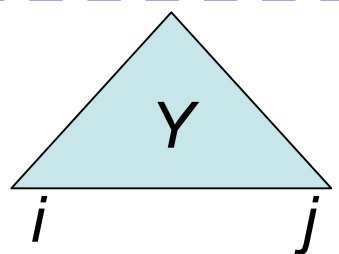


$X \rightarrow w_i$



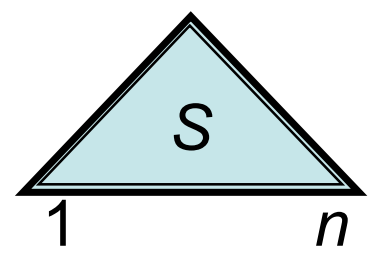
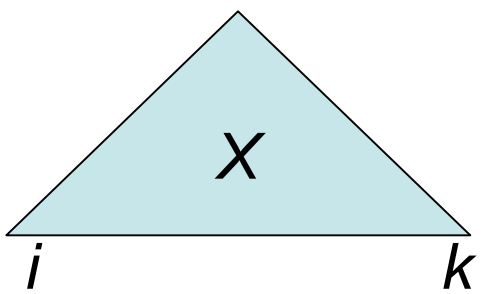
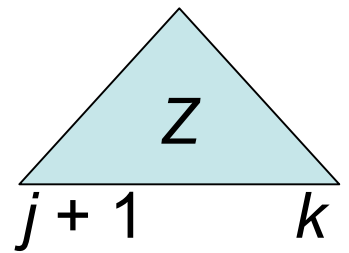
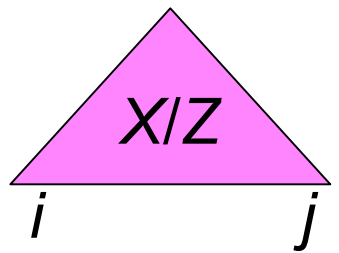
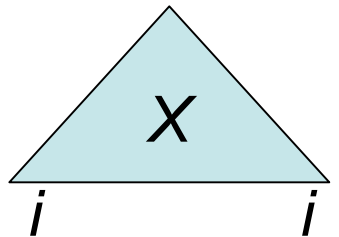
$\{X, Y, Z, i, j\}$ CKY

$X \rightarrow YZ$



Runtime to instantiate all inferences: $O(|N|^3 n^2)$

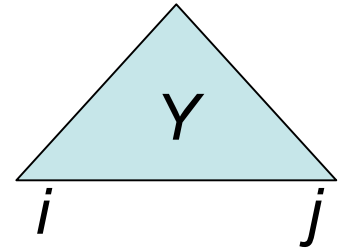
$X \rightarrow w_i$



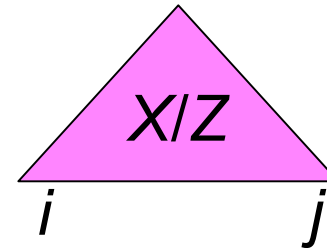
CKY

$\{X, Y, i, j, k\}$

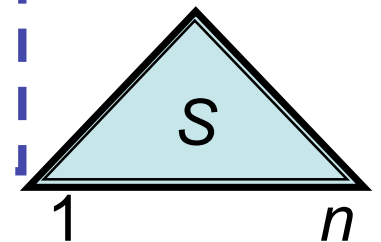
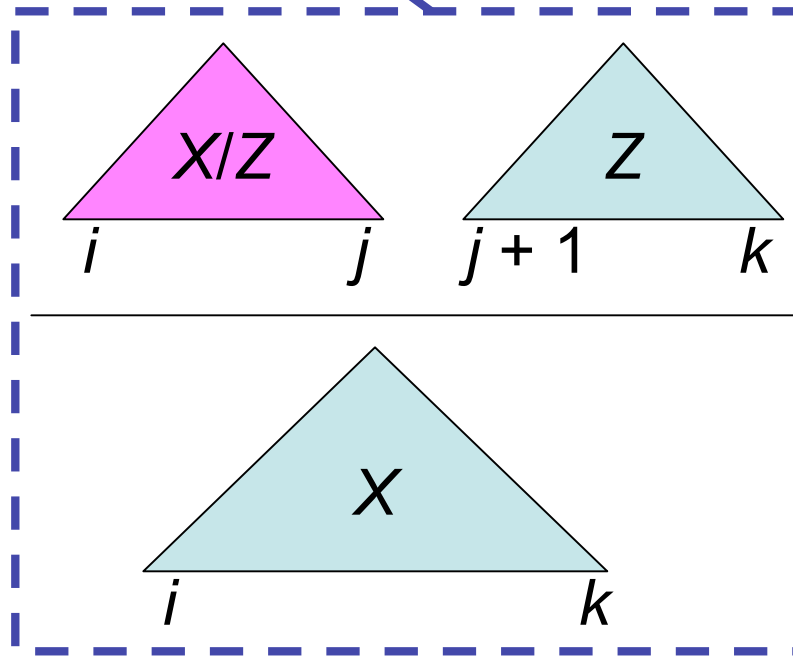
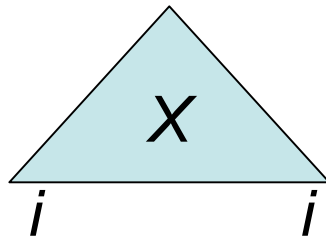
$X \rightarrow YZ$



Runtime to instantiate
all inferences: $O(|N|^2n^3)$



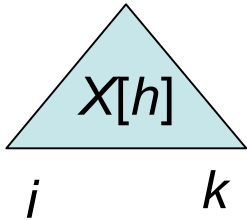
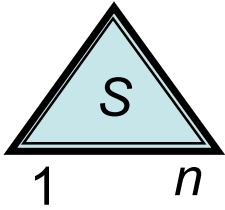
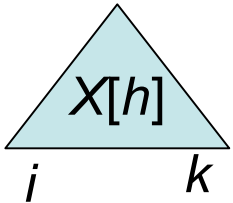
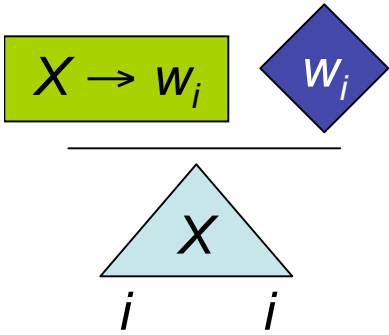
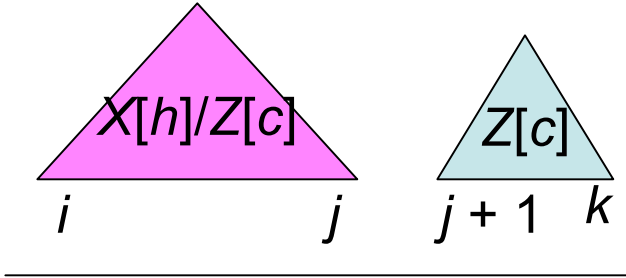
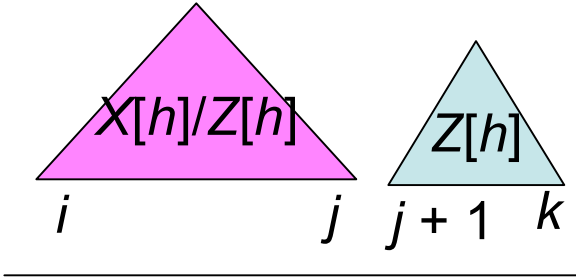
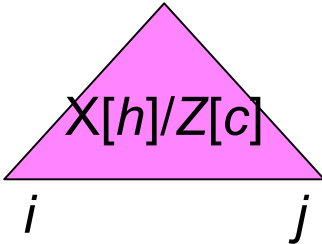
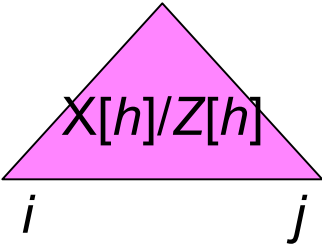
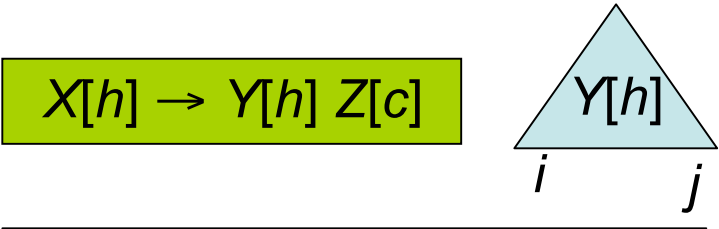
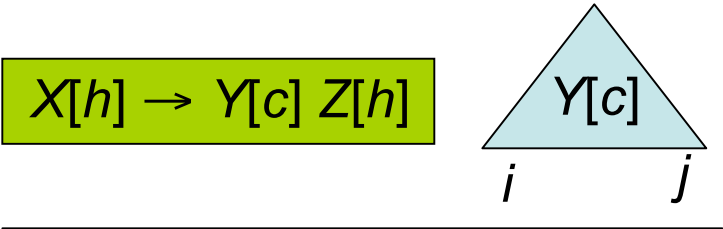
$X \rightarrow w_i$



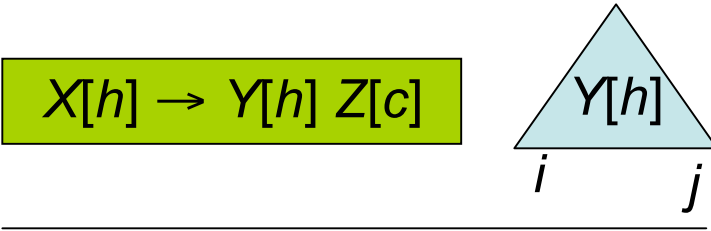
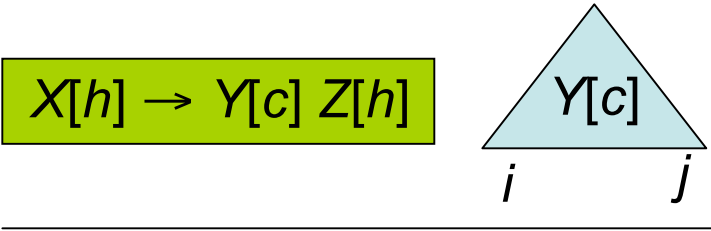
CKY

- Runtime reduced from $O(|N|^3n^3)$ to $O(|N|^2n^3 + |N|^3n^2)$.
- Related example: lexicalized PCFG parsing (Eisner and Satta, 1999).

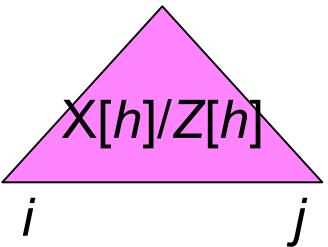
Lexicalized CKY



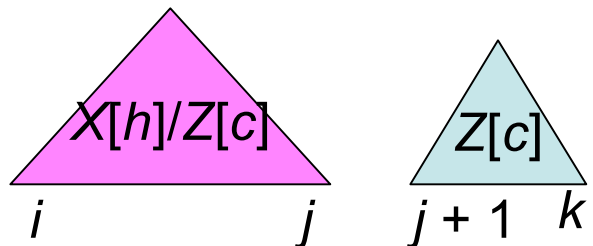
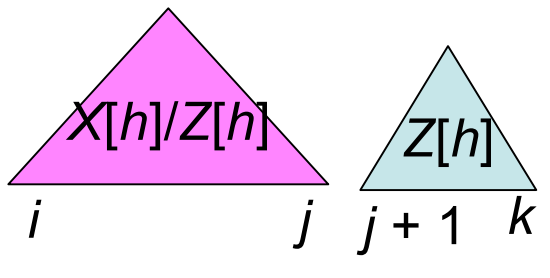
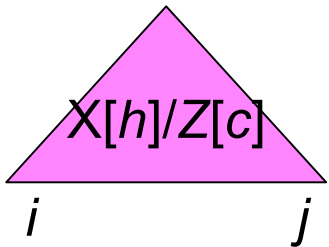
Lexicalized CKY



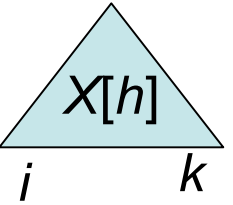
$O(|N|^3 n^4)$



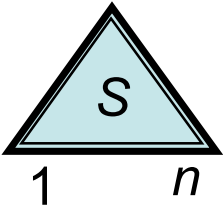
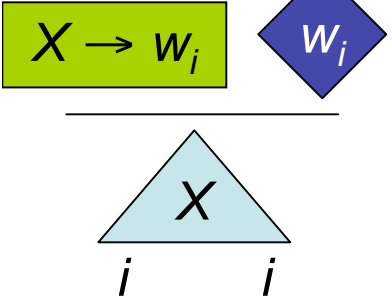
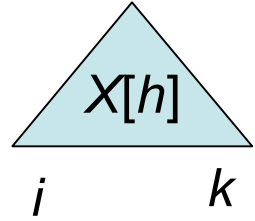
$O(|N|^3 n^4)$



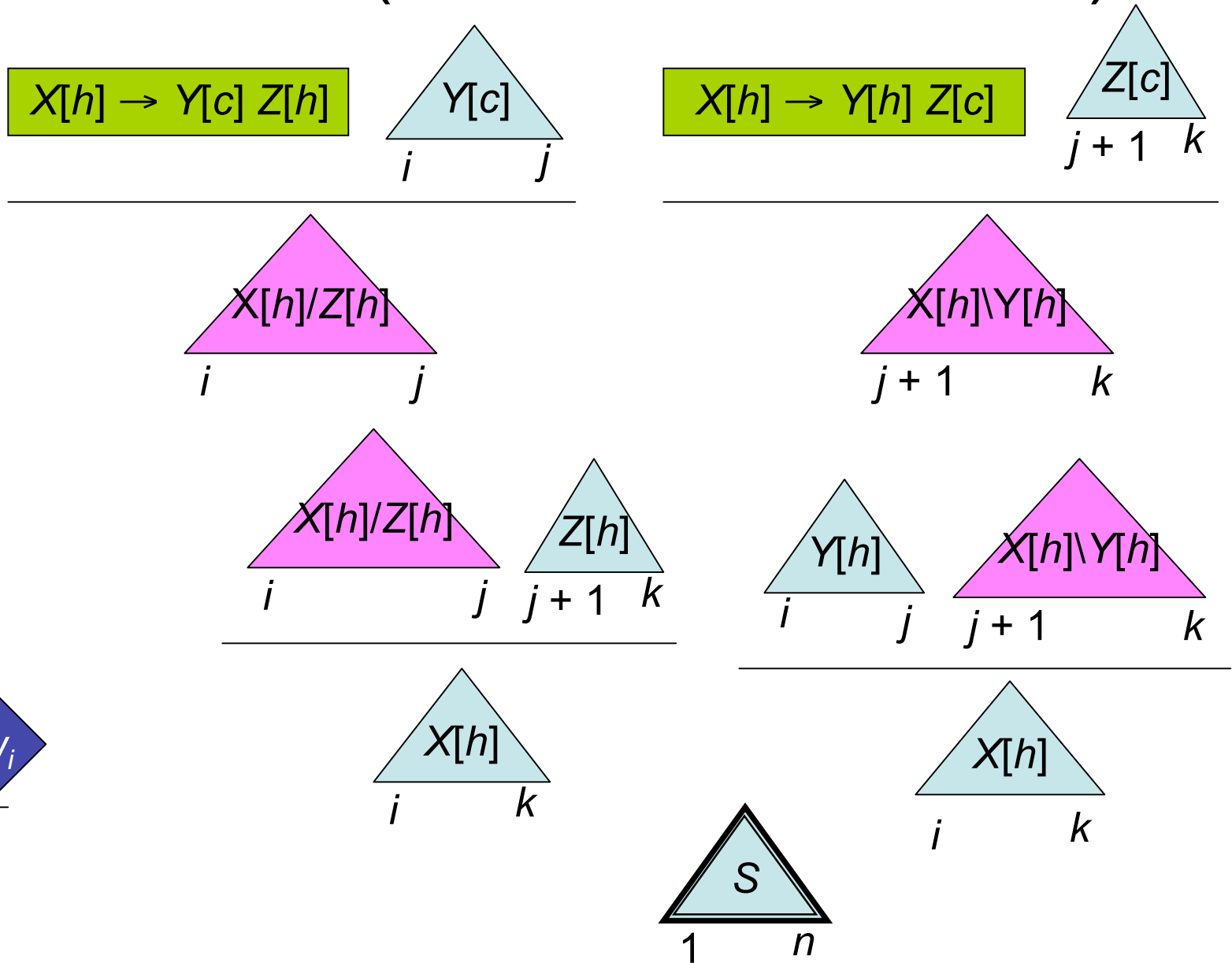
$O(|N|^2 n^4)$



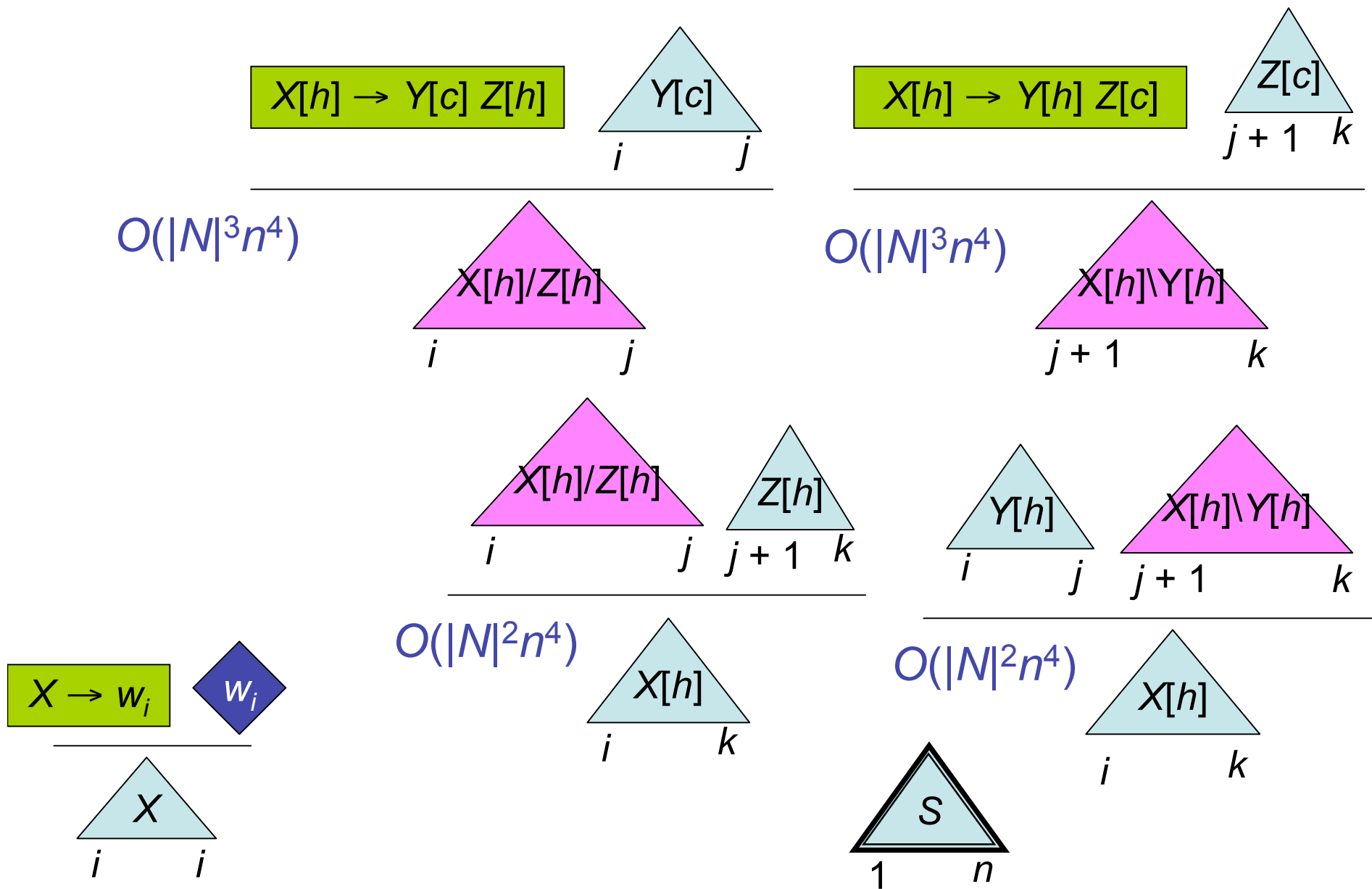
$O(|N|^2 n^5)$



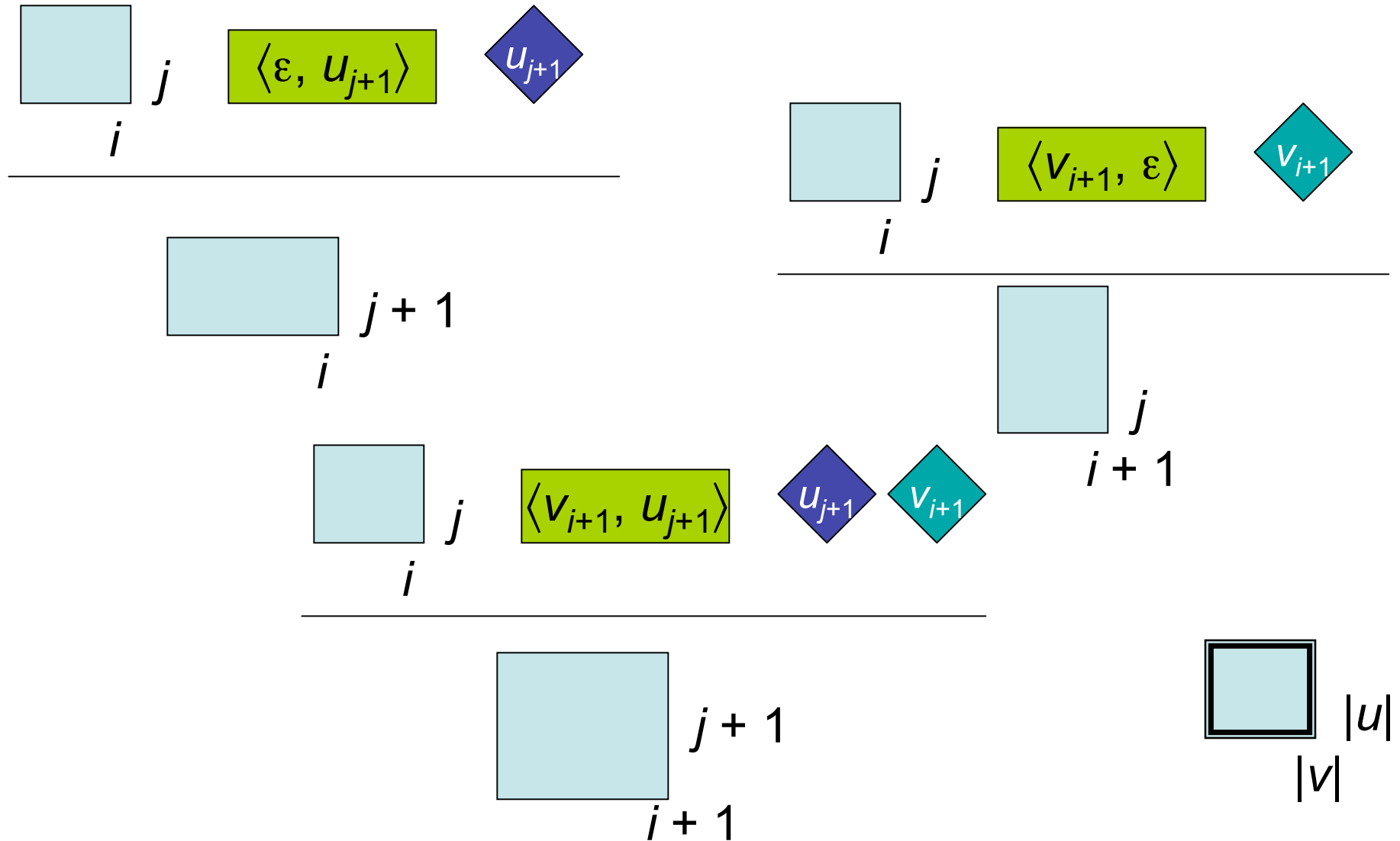
Lexicalized CKY (Eisner & Satta, 1999)



Lexicalized CKY (Eisner & Satta, 1999)

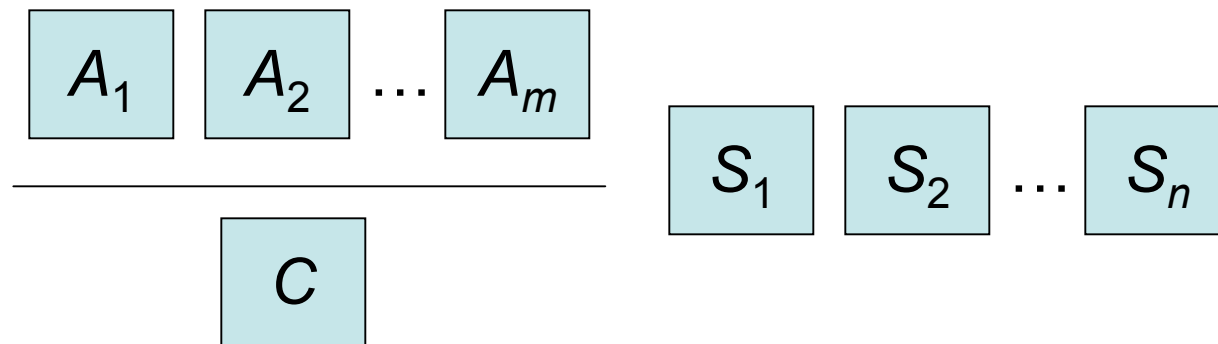


String Alignment



What about weights?

- Goodman (1999): add weights.
- Not limited to just probabilities!
 - Semiring weights.



$$V(C) \oplus = \begin{cases} \bigotimes_{i=1}^m V(A_i) & \text{if } \bigwedge_{j=1}^n [V(S_j) \neq \mathbf{0}] \\ \mathbf{0} & \text{otherwise} \end{cases}$$

Semirings

			boolean	Viterbi
set	\mathbb{V}	-	{false, true}	\mathbb{R}_+
plus	\oplus	associative and commutative	\vee	max
times	\otimes	associative, distributes over \oplus	\wedge	\times
zero	0	\oplus -identity: $x \oplus \mathbf{0} = x$	false	0
one	1	\otimes -identity: $x \otimes \mathbf{1} = x$	true	1

Semirings

	boolean	Viterbi	inside	counting	derivation forest
set	{false, true}	\mathbb{R}_+	\mathbb{R}_+	\mathbb{N}_+	2^E
plus	\vee	max	+	+	\cup
times	\wedge	\times	\times	\times	concatenation
zero	false	0	0	0	\emptyset
one	true	1	1	1	$\{\langle \rangle\}$

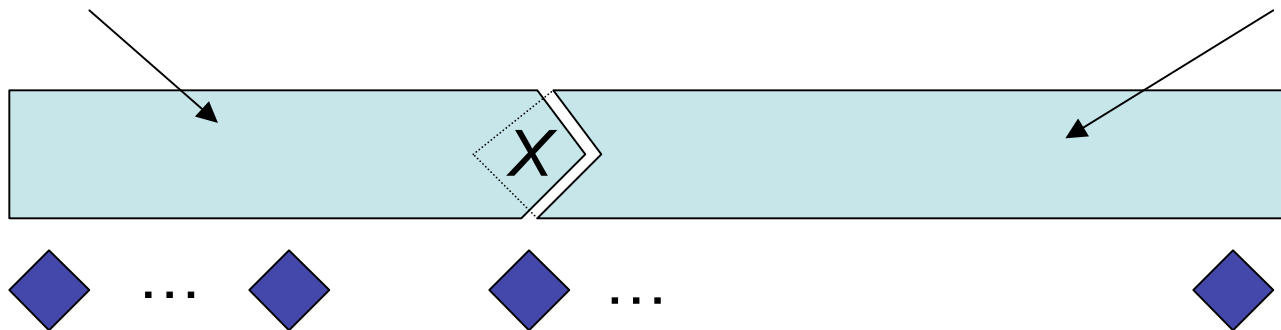
Goodman (1999)

- Viterbi-derivation and Viterbi- n -best semirings defined, as well.
- **Reverse** values can be computed in the same framework!

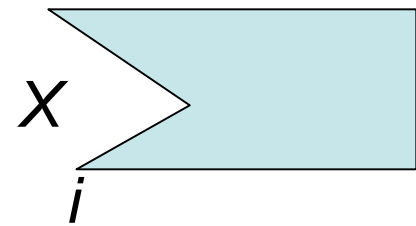
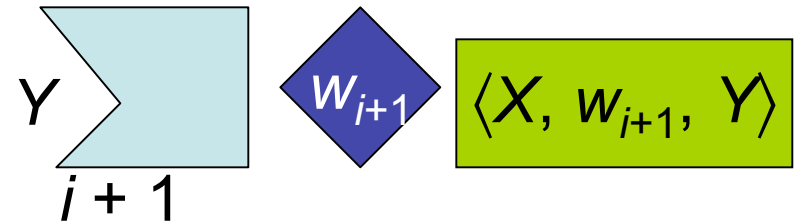
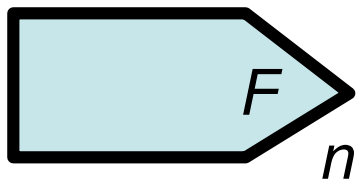
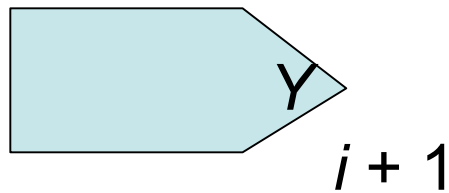
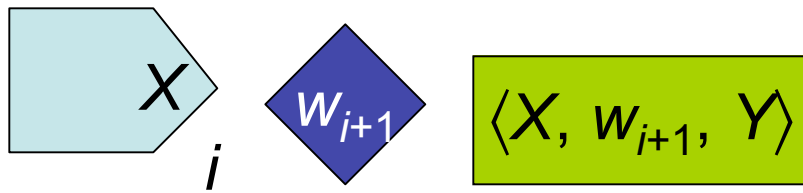
Forward and Backward Weights

sum over all partial structures

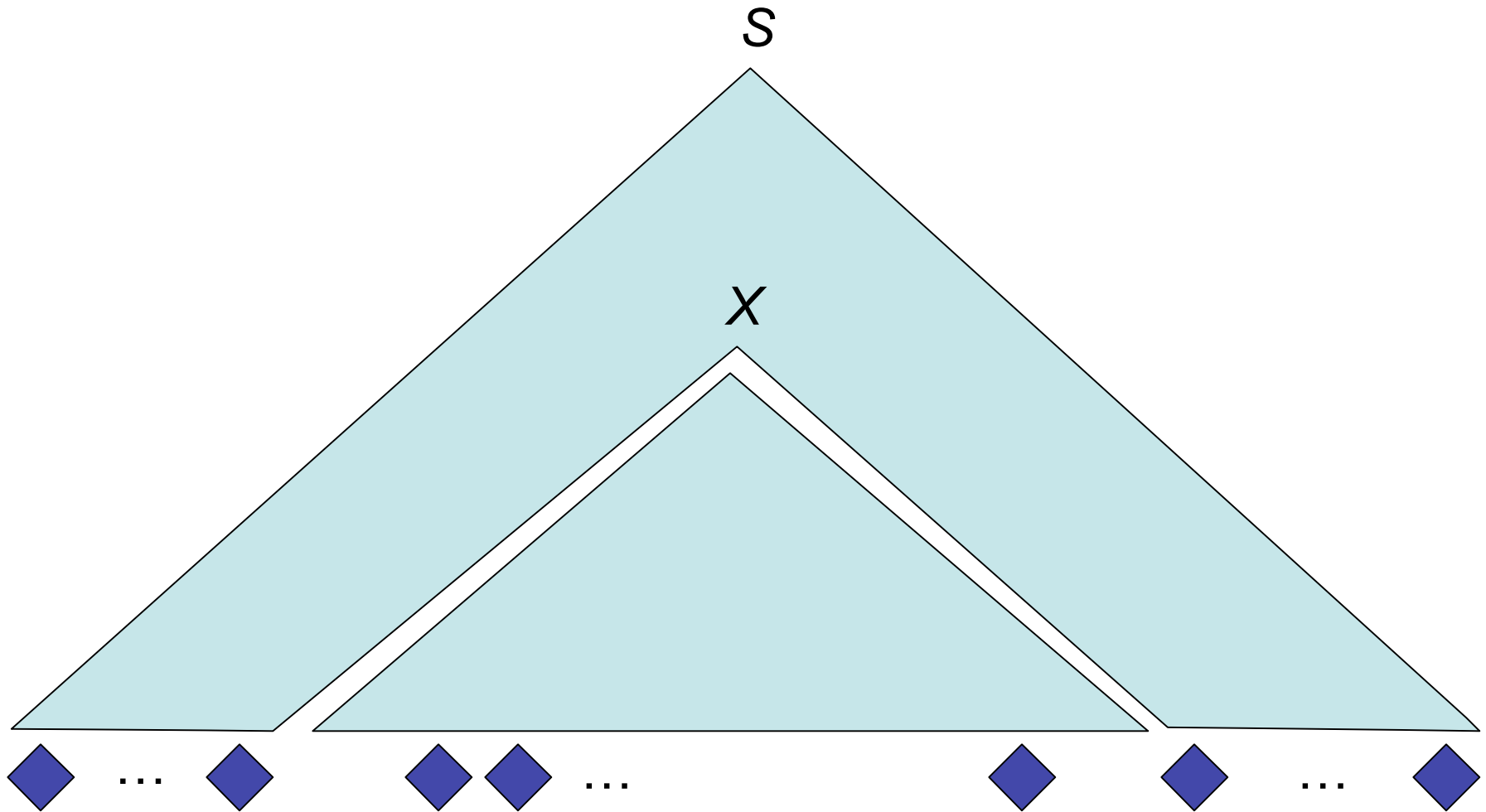
sum over all partial structures



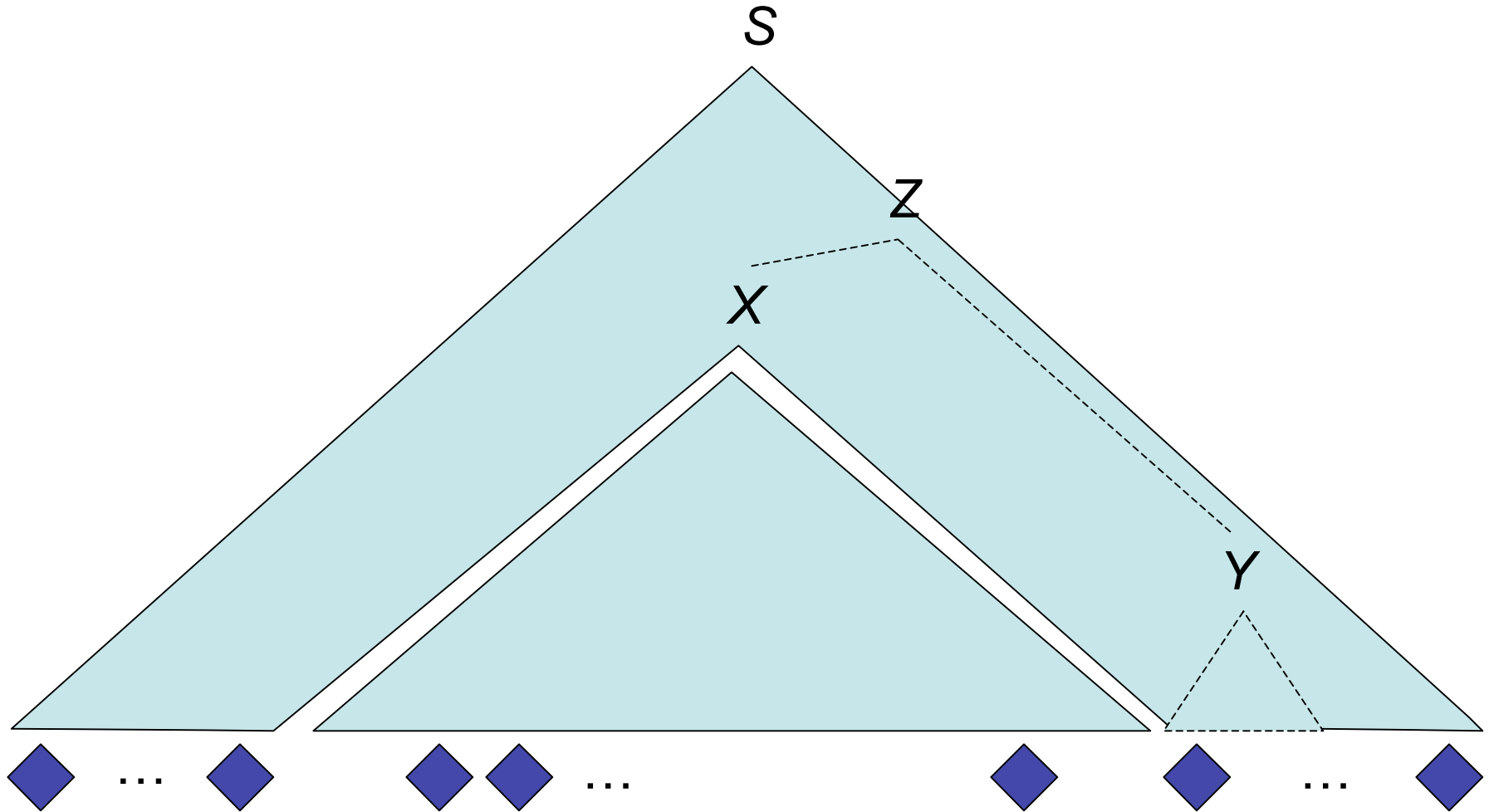
Forward and Backward Algorithms



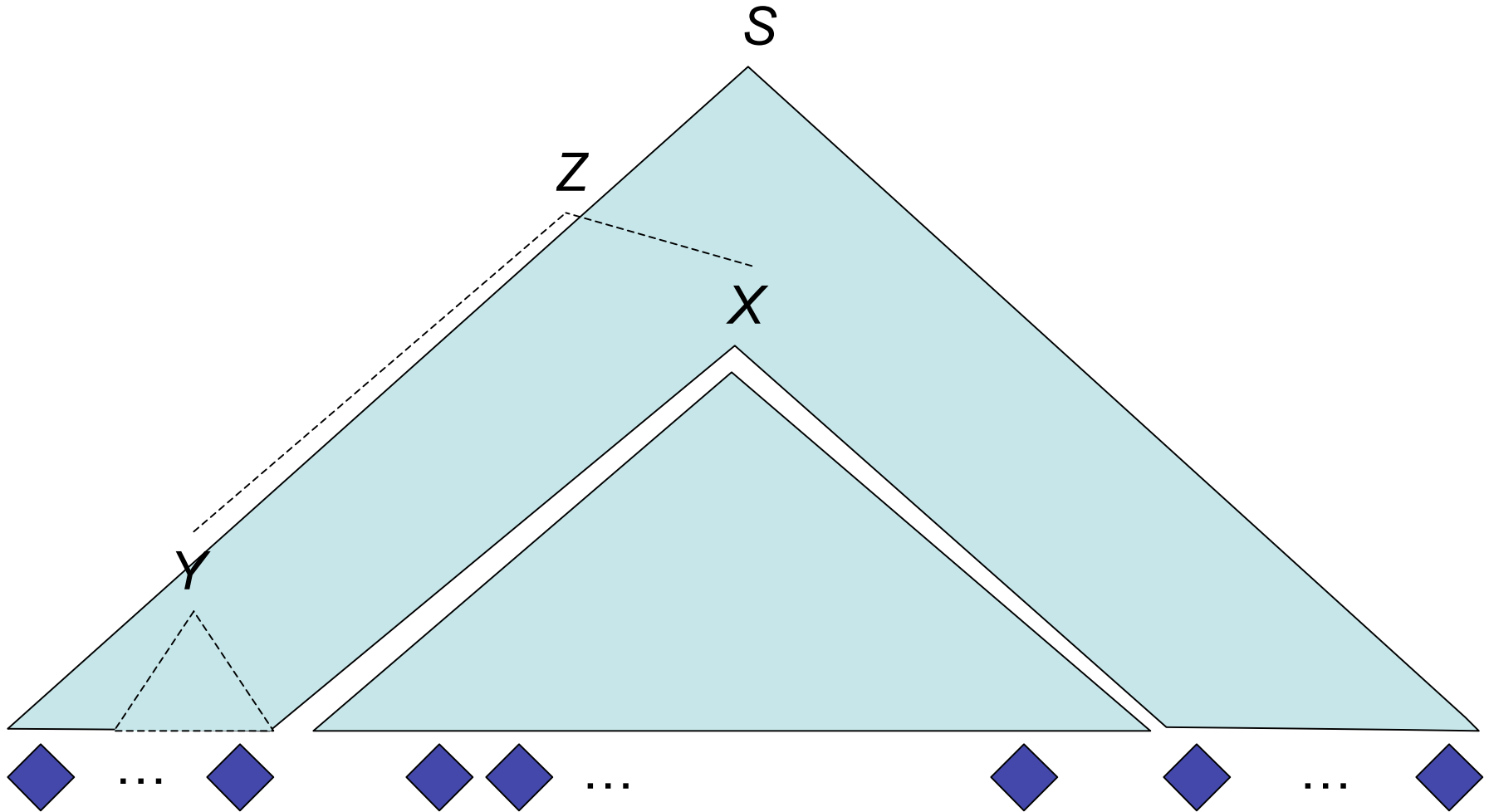
Inside and Outside Weights



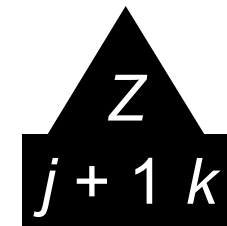
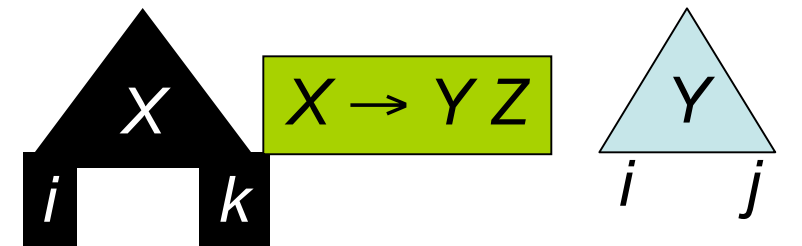
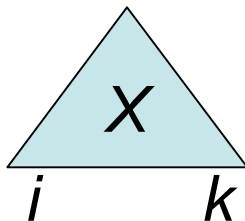
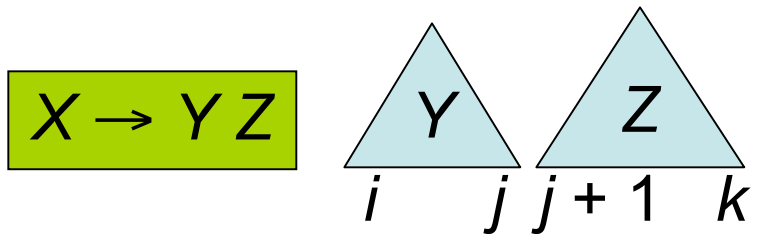
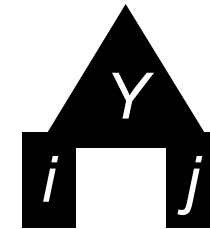
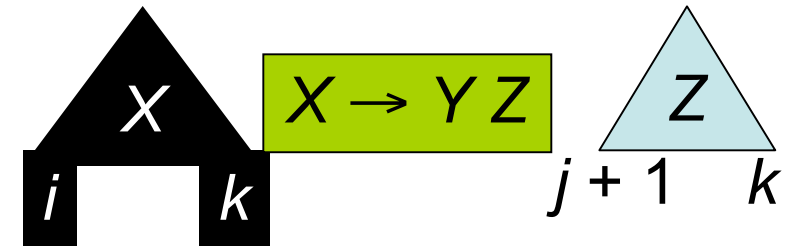
Inside and Outside Weights



Inside and Outside Weights



Outside (CKY)



Inside and Outside Algorithms

```
float chart[1..n, 1..|N|, 1..n+1] := 0;
for s := 1 to n /* start position */
  for each rule  $A \rightarrow w_s \in R$ 
    chart[s, A, s+1] :=  $P(A \rightarrow w_s)$ ;
for l := 2 to n /* length, shortest to longest */
  for s := 1 to n-l+1 /* start position */
    for t := 1 to l-1 /* split length */
      for each rule  $A \rightarrow BC \in R$ 
        chart[s, A, s+l] := chart[s, A, s+l] +
          (chart[s, B, s+t]  $\times$  chart[s+t, C, s+l]  $\times$   $P(A \rightarrow BC)$ );
return chart[1, S, n+1];
```

```
float outside[1..n, 1..|N|, 1..n+1] := 0;
outside[1, S, n+1] := 1;
for l := n down to 2 /* length, longest to shortest */
  for s := 1 to n-l+1 /* start position */
    for t := 1 to l-1 /* split length */
      for each rule  $A \rightarrow BC \in R$ 
        outside[s, B, s+t] := outside[s, B, s+t] +
          (outside[s, A, s+l]  $\times$  inside[s+t, C, s+l]  $\times$   $P(A \rightarrow BC)$ );
        outside[s+t, C, s+l] := outside[s+t, C, s+l] +
          (outside[s, A, s+l]  $\times$  inside[s, B, s+t]  $\times$   $P(A \rightarrow BC)$ );
```

Beyond Goodman (1999)

- Algorithm: carry out deduction to build the chart (i.e., fill in items with nonzero value); then compute their values.
 - Tough part: **efficient** ordering of items.
 - For Forward/Viterbi: order by position
 - For CKY: order by width
 - In general?
- Need efficient **execution strategy** for **arbitrary** programs.
 - Key idea: avoid unnecessary work and repropagation.

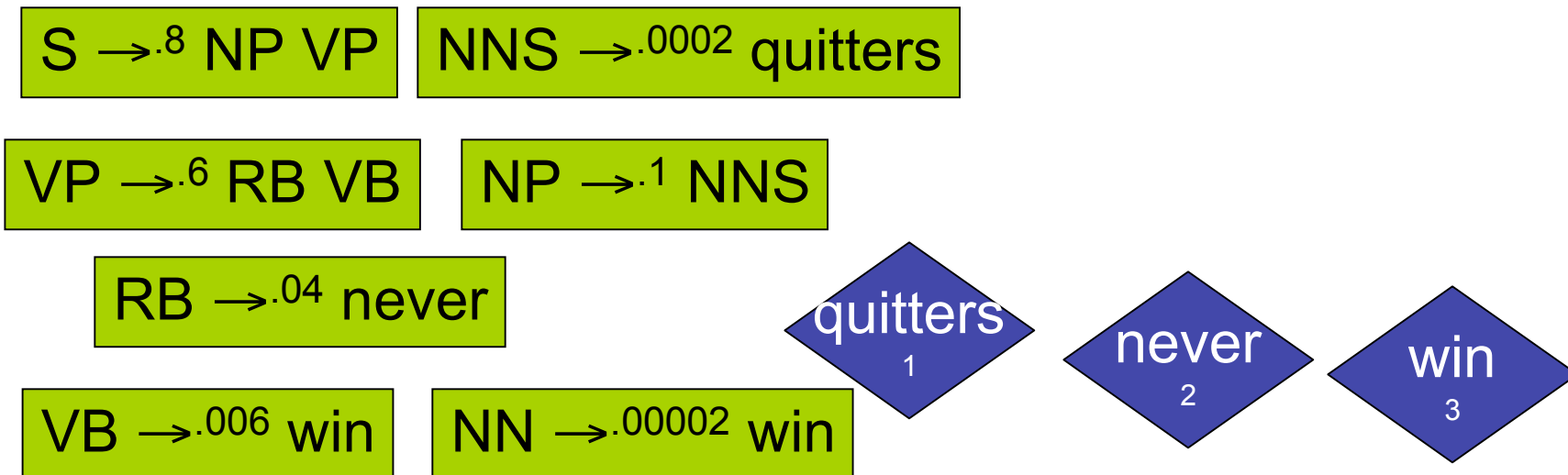
Unnecessary Work

- If you only want the best derivation, you don't want to build items that aren't in it!
- But you don't know which items to build until you have the best parse.
- Key idea: **agenda**.
 - Order **updates** to items' weights.
 - Roughly analogous to trading depth and breadth in **search**.
- Note: for exact inside/outside, all of the work **is** necessary!

Repropagation

- Suppose $(NP, 4, 7)$ currently has a weight of 0.3, constructed by $(DT, 4, 5) \otimes (NP, 5, 7)$.
- Now suppose we find that **a better** way to build $(NP, 4, 7)$: $(DT, 4, 5) \otimes (NNP, 5, 6) \otimes (NNP, 6, 7)$ with value 0.31.
- Maybe now we have a better way to build $(VP, 3, 7)$! (Or anything else that used $(NP, 4, 7)$).
- Have to re-build all of those consequents, and compare again, and recursively repropagate to consequents of any item whose value changes.
- May not be $O(n^3)$ anymore!

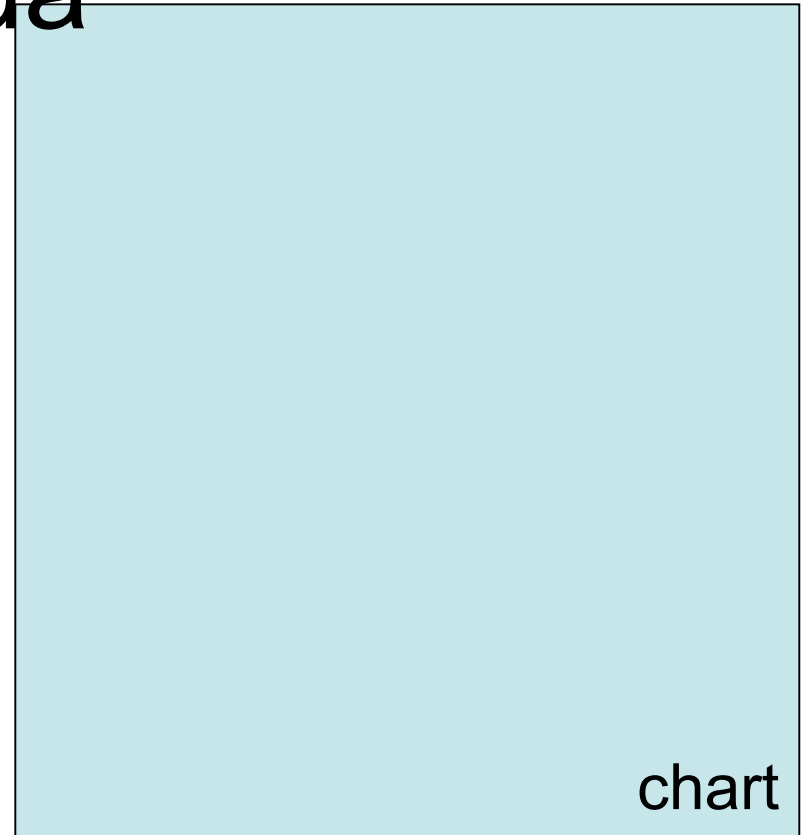
Agenda



...

Agenda

quitters
1



S \rightarrow .8 NP VP NNS \rightarrow .0002 quitters

VP \rightarrow .6 RB VB NP \rightarrow .1 NNS

RB \rightarrow .04 never

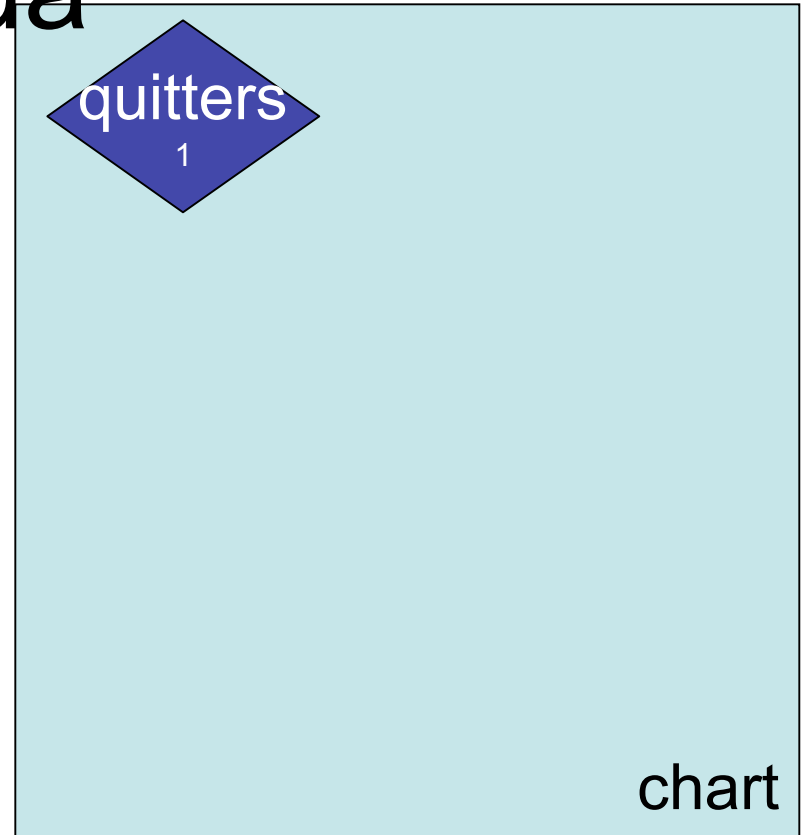
VB \rightarrow .006 win NN \rightarrow .00002 win

...

never
2

win
3

Agenda



S \rightarrow .8 NP VP NNS \rightarrow .0002 quitters

VP \rightarrow .6 RB VB NP \rightarrow .1 NNS

RB \rightarrow .04 never

VB \rightarrow .006 win NN \rightarrow .00002 win

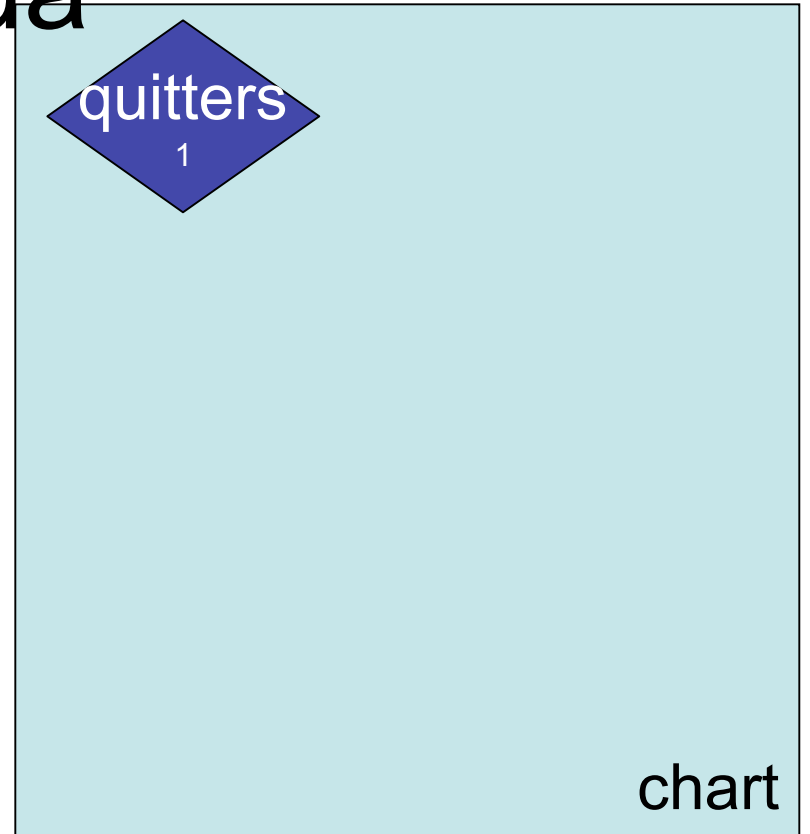
...



Agenda

never
2

quitters
1



S \rightarrow .8 NP VP NNS \rightarrow .0002 quitters

VP \rightarrow .6 RB VB NP \rightarrow .1 NNS

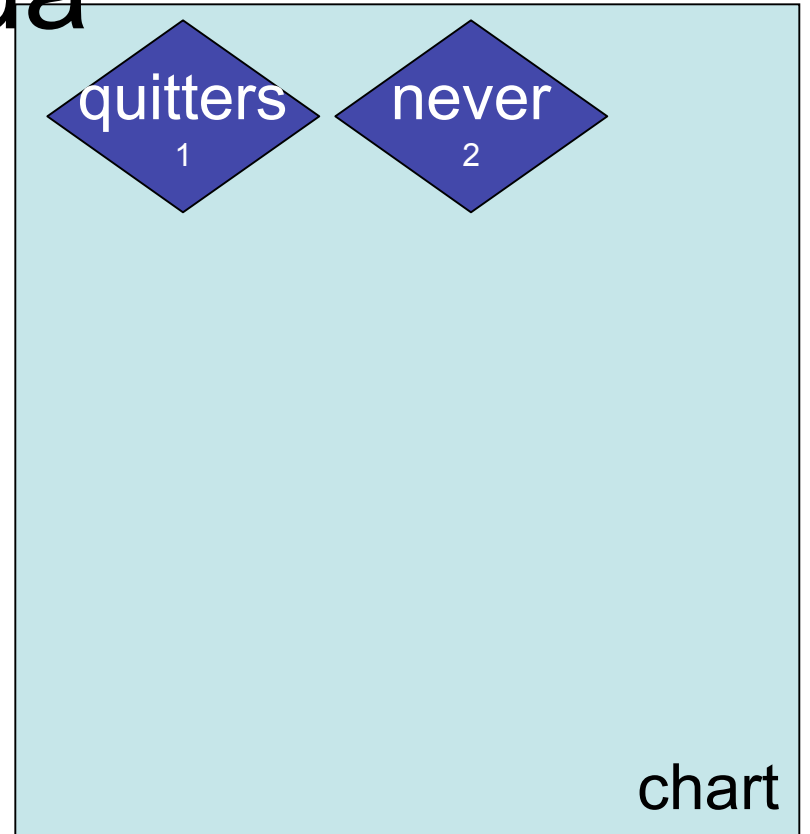
RB \rightarrow .04 never

VB \rightarrow .006 win NN \rightarrow .00002 win

...

win
3

Agenda



S \rightarrow .8 NP VP

NNS \rightarrow .0002 quitters

VP \rightarrow .6 RB VB

NP \rightarrow .1 NNS

RB \rightarrow .04 never

VB \rightarrow .006 win

NN \rightarrow .00002 win

...



Agenda

win
3

quitters
1

never
2

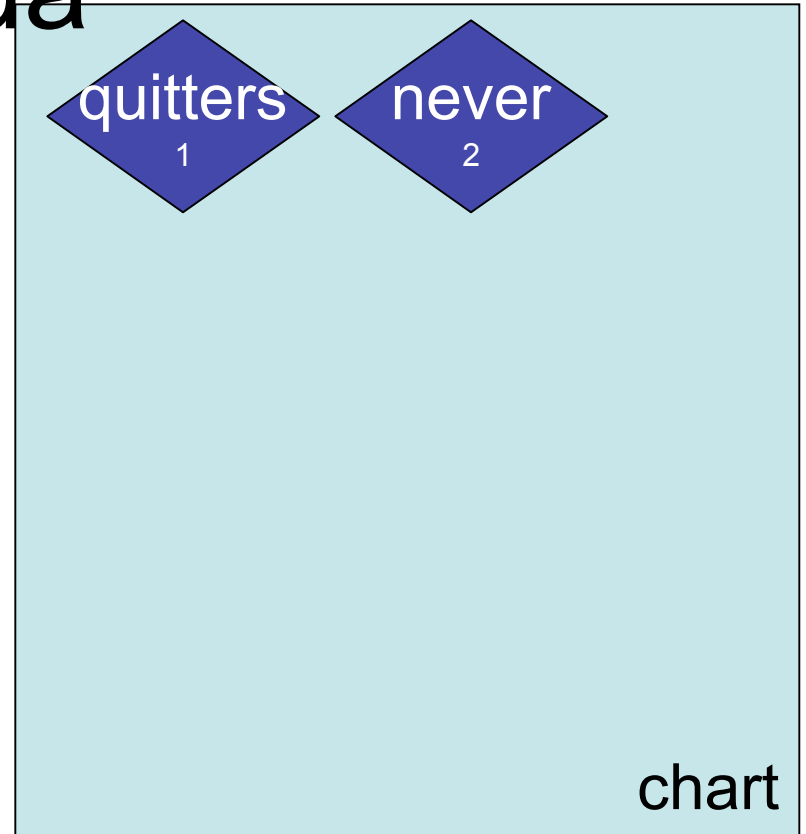
S \rightarrow .8 NP VP NNS \rightarrow .0002 quitters

VP \rightarrow .6 RB VB NP \rightarrow .1 NNS

RB \rightarrow .04 never

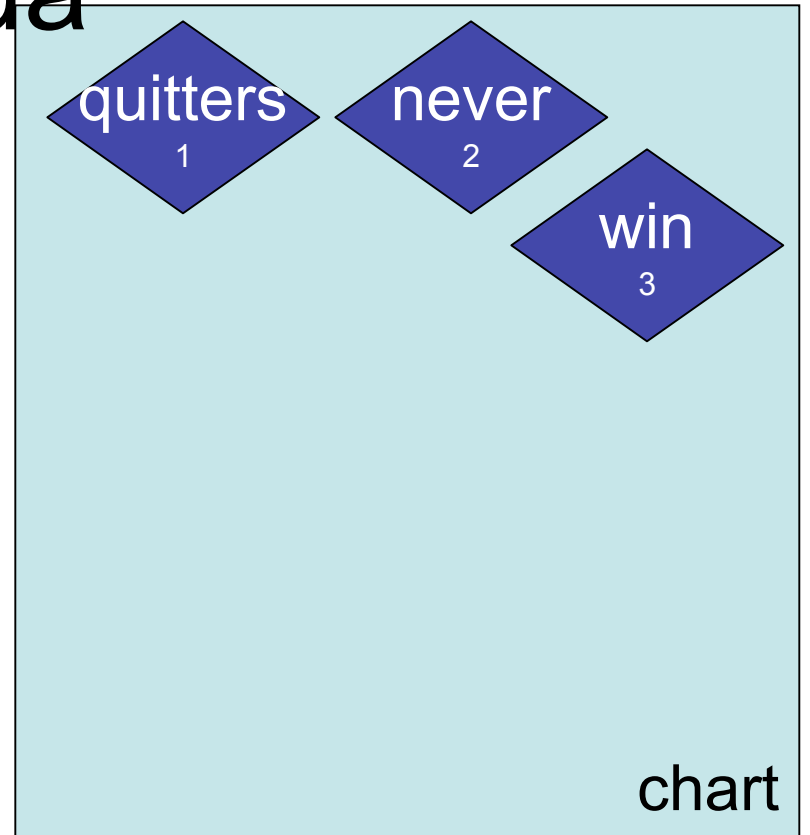
VB \rightarrow .006 win NN \rightarrow .00002 win

...



chart

Agenda



S \rightarrow .8 NP VP NNS \rightarrow .0002 quitters

VP \rightarrow .6 RB VB NP \rightarrow .1 NNS

RB \rightarrow .04 never

VB \rightarrow .006 win NN \rightarrow .00002 win

...

Agenda

S \rightarrow .8 NP VP

NNS \rightarrow .0002 quitters

VP \rightarrow .6 RB VB

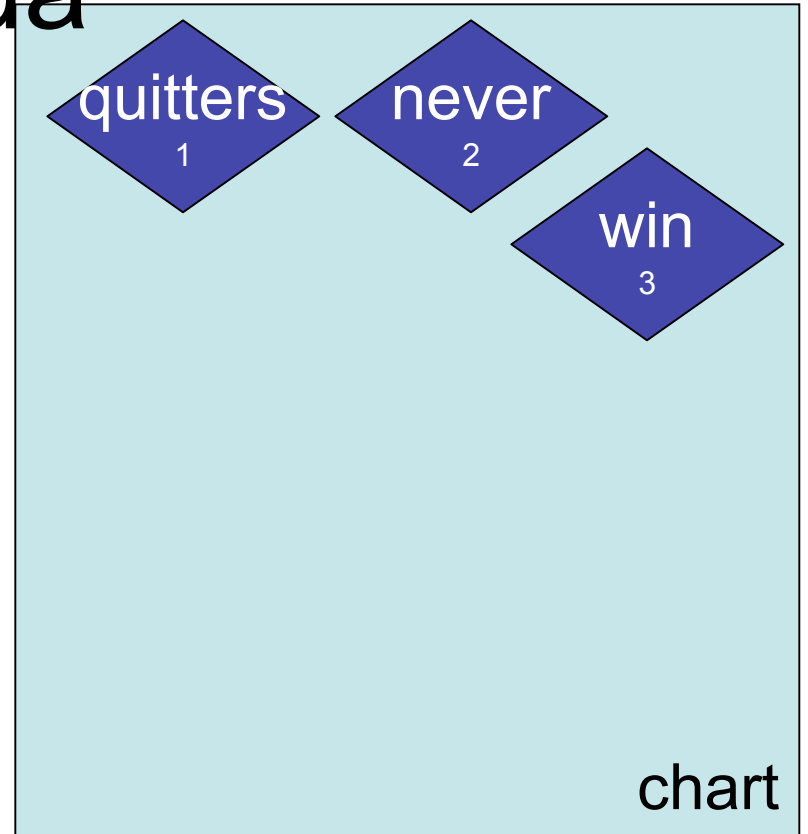
NP \rightarrow .1 NNS

RB \rightarrow .04 never

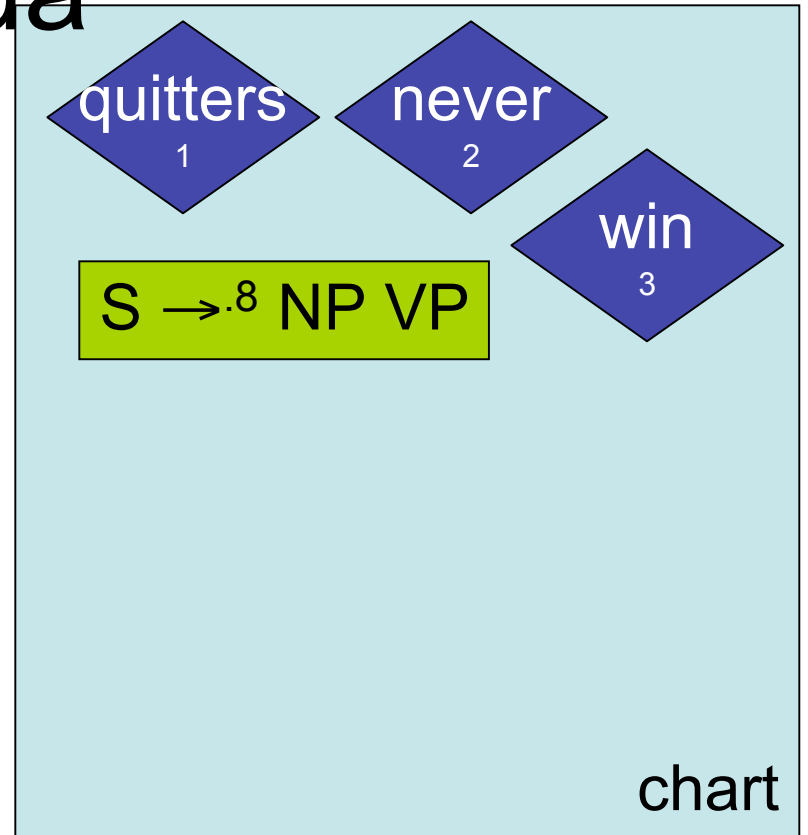
VB \rightarrow .006 win

NN \rightarrow .00002 win

...



Agenda



NNS \rightarrow .0002 quitters

VP \rightarrow .6 RB VB

NP \rightarrow .1 NNS

RB \rightarrow .04 never

VB \rightarrow .006 win

NN \rightarrow .00002 win

...

Agenda

VP \rightarrow .6 RB VB

NNS \rightarrow .0002 quitters

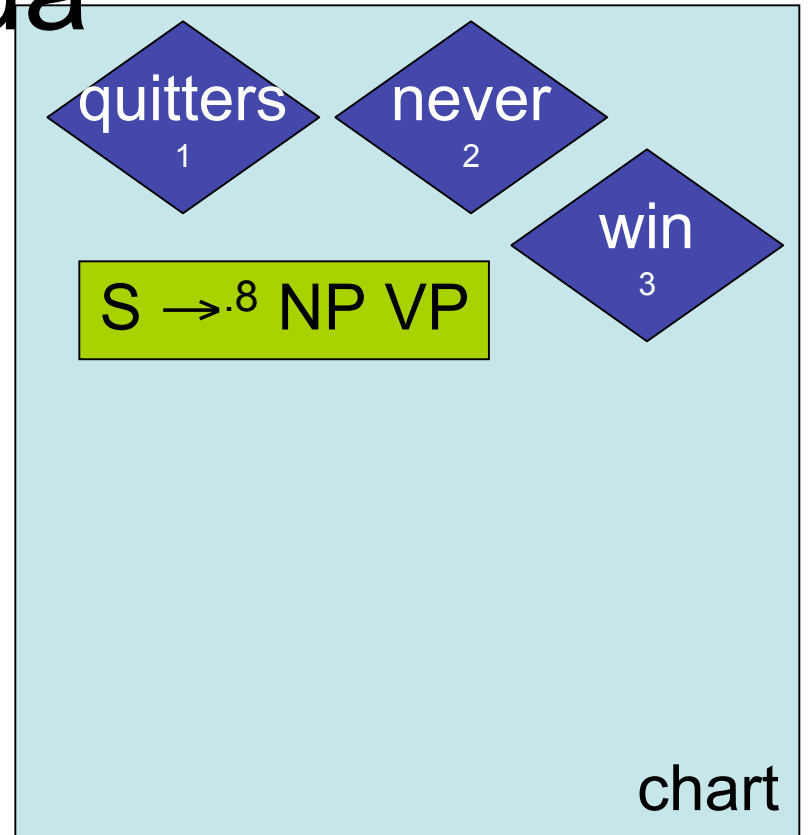
NP \rightarrow .1 NNS

RB \rightarrow .04 never

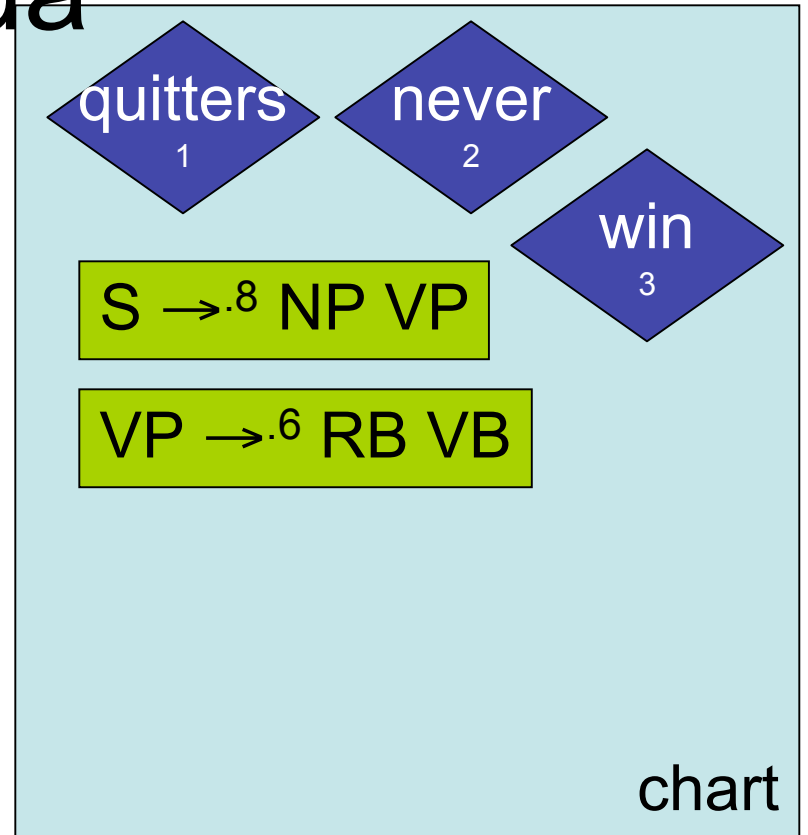
VB \rightarrow .006 win

NN \rightarrow .00002 win

...



Agenda



NNS → .0002 quitters

NP → .1 NNS

RB → .04 never

VB → .006 win

NN → .00002 win

...

Agenda

NP \rightarrow .1 NNS

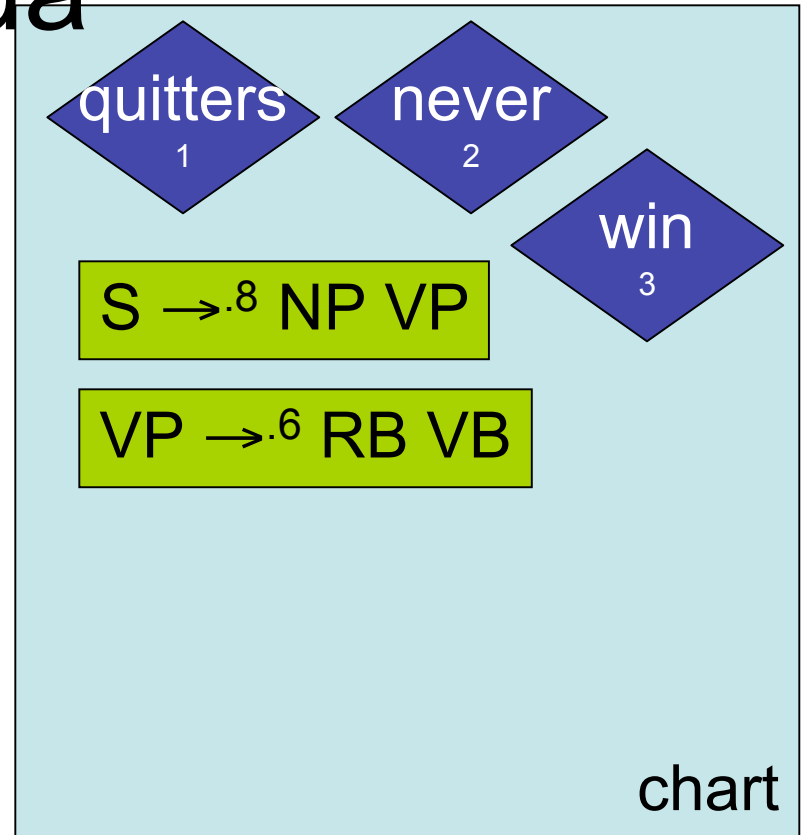
NNS \rightarrow .0002 quitters

RB \rightarrow .04 never

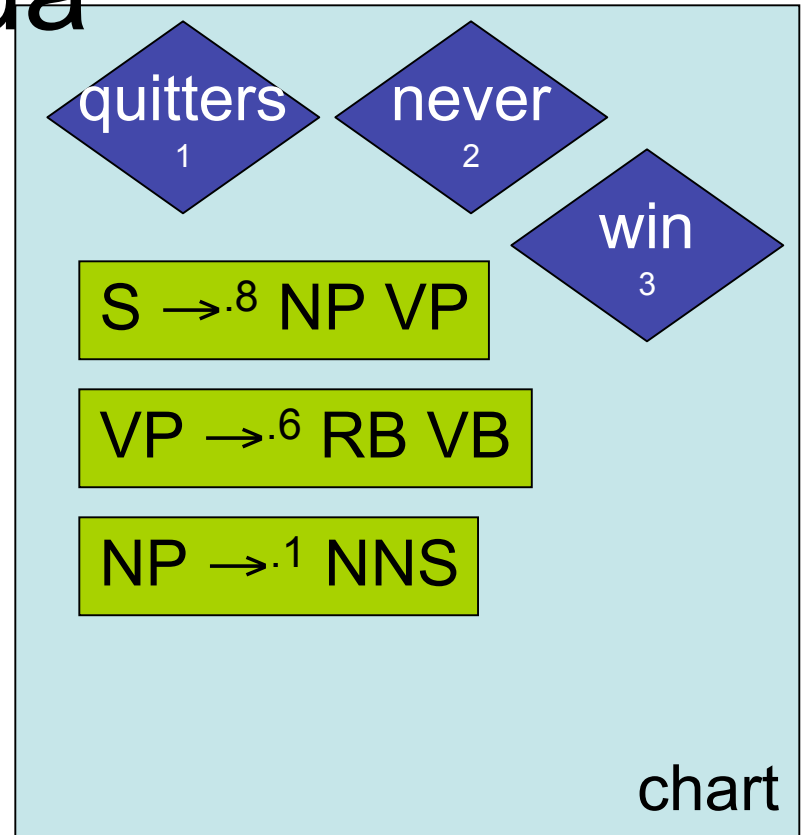
VB \rightarrow .006 win

NN \rightarrow .00002 win

...



Agenda



NNS → .0002 quitters

RB → .04 never

VB → .006 win

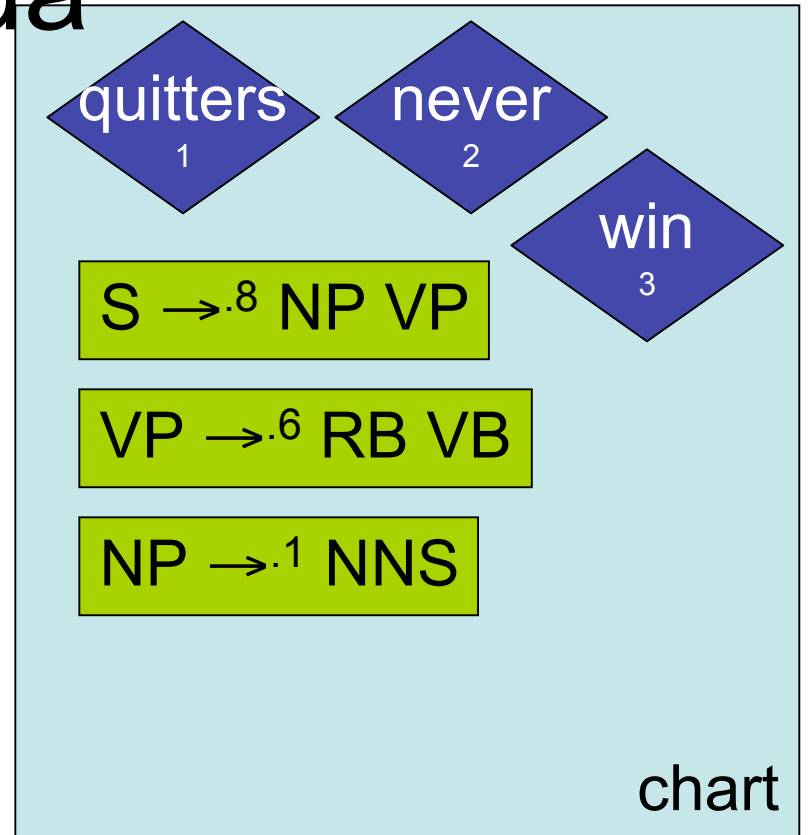
NN → .00002 win

...

Agenda

RB \rightarrow .04 never

NNS \rightarrow .0002 quitters

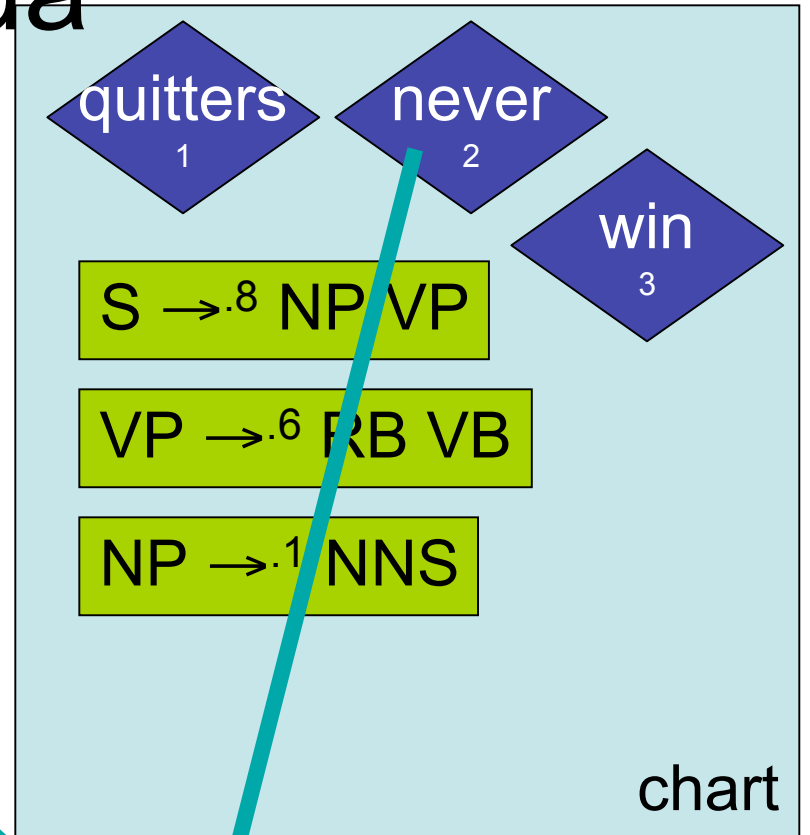


VB \rightarrow .006 win

NN \rightarrow .00002 win

...

Agenda



RB → .04 never

NNS → .0002 quitters

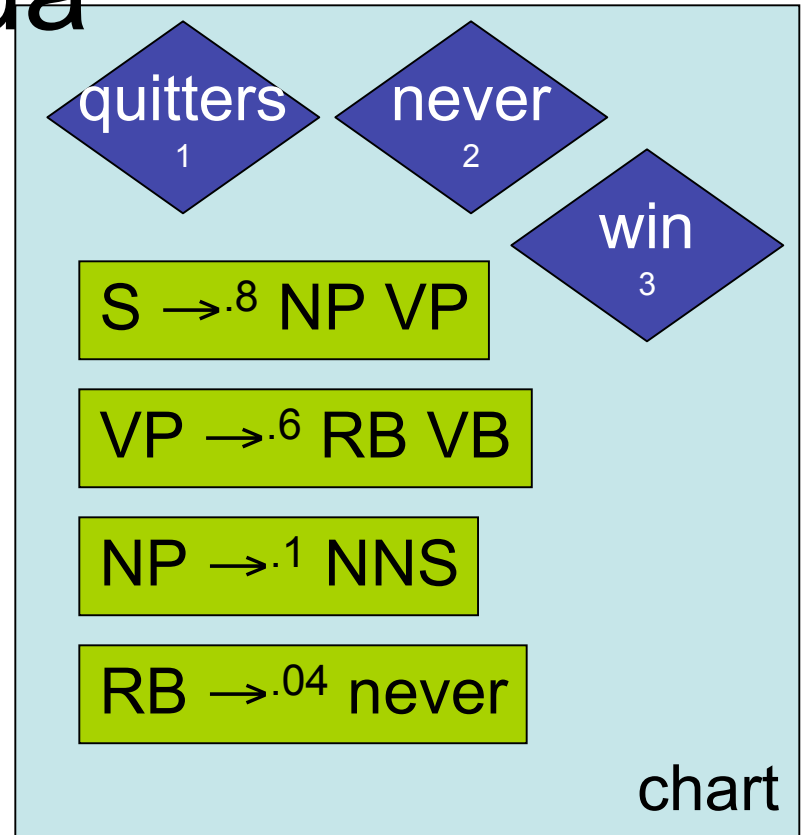
VB → .006 win

NN → .00002 win

...

${}_2\text{RB}_2$
.04

Agenda



NNS \rightarrow .0002 quitters

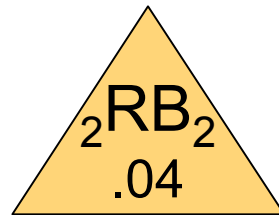
${}^2\text{RB}_2$
.04

VB \rightarrow .006 win

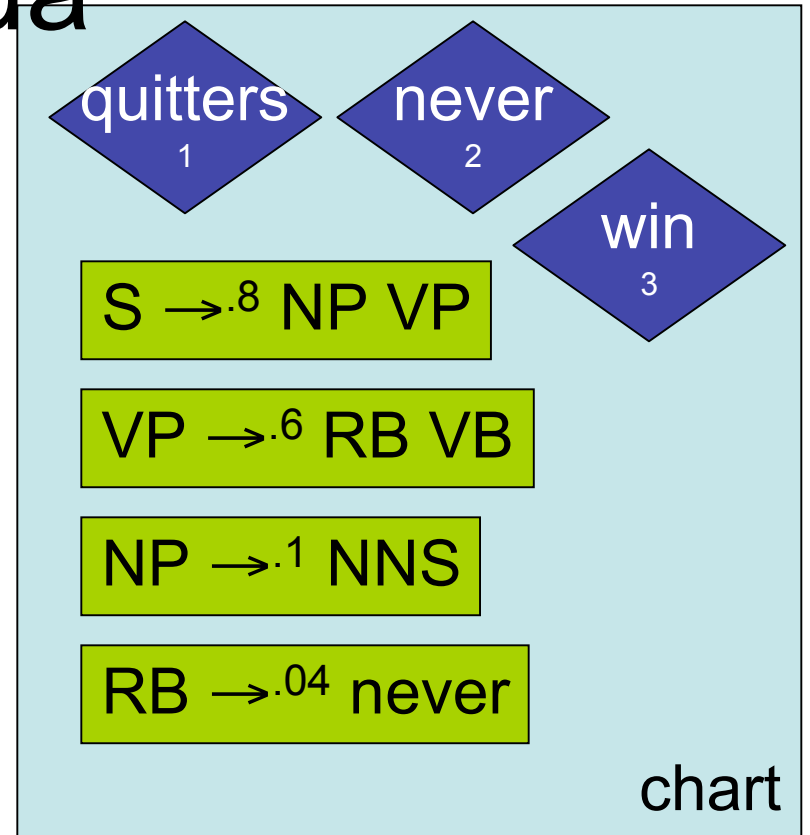
NN \rightarrow .00002 win

...

Agenda



$\text{NNS} \rightarrow .0002 \text{ quitters}$

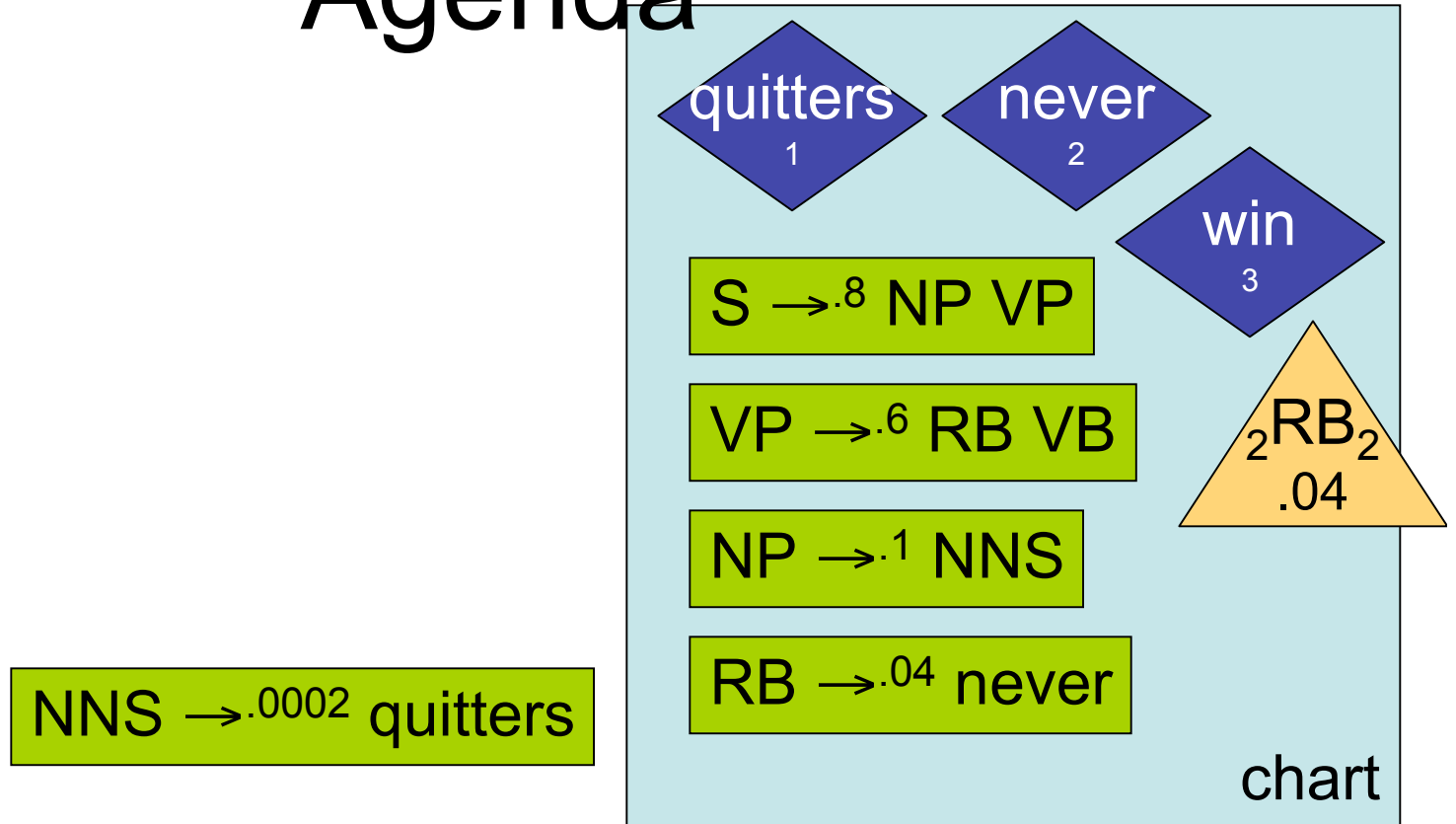


$\text{VB} \rightarrow .006 \text{ win}$

$\text{NN} \rightarrow .00002 \text{ win}$

...

Agenda



VB → .006 win

NN → .00002 win

...

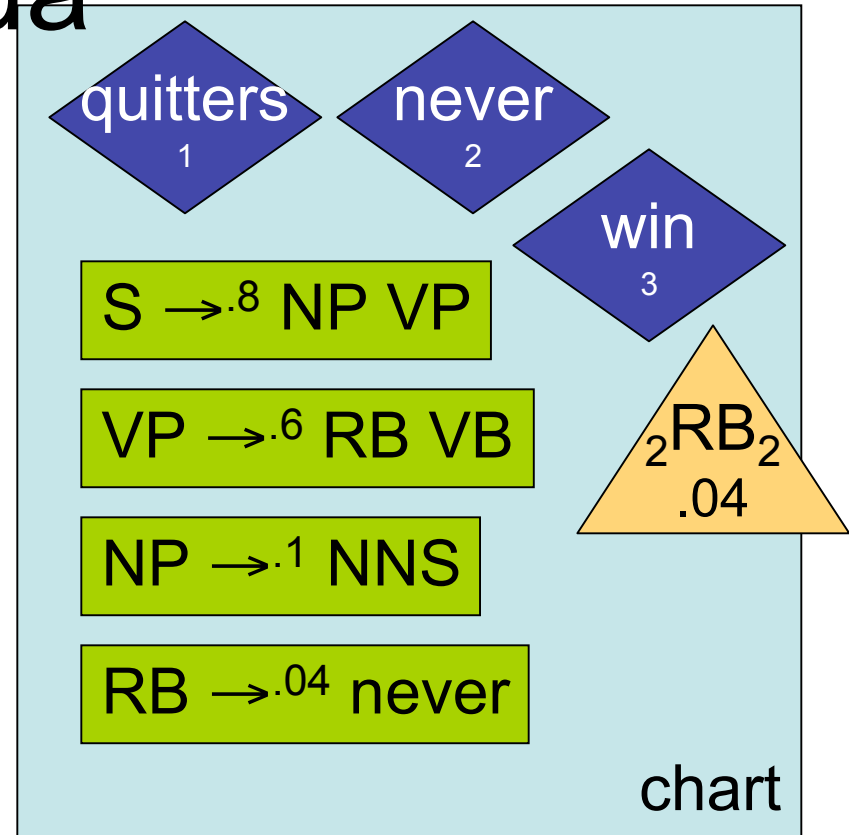
Agenda

VB \rightarrow .006 win

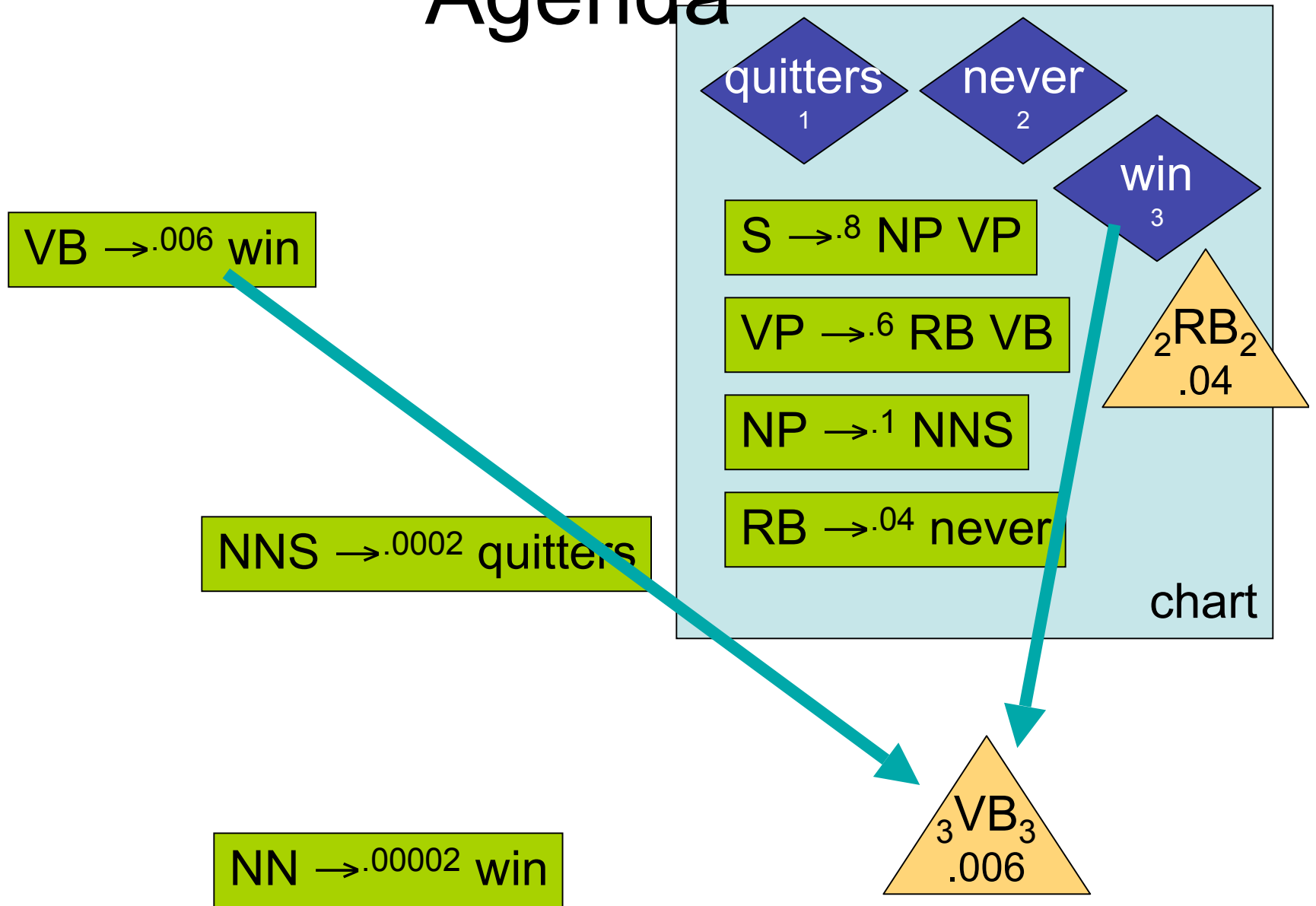
NNS \rightarrow .0002 quitters

NN \rightarrow .00002 win

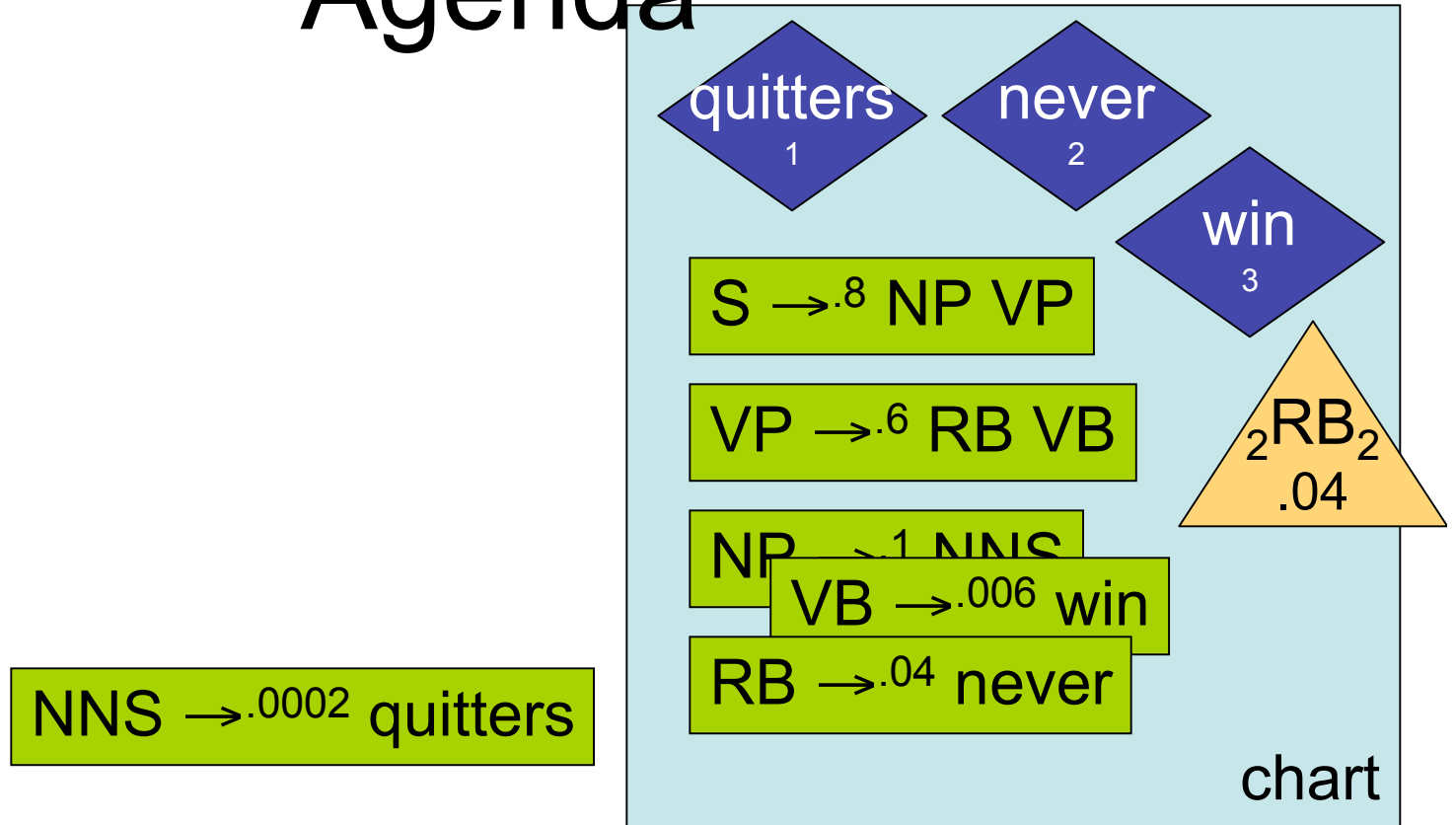
...



Agenda



Agenda

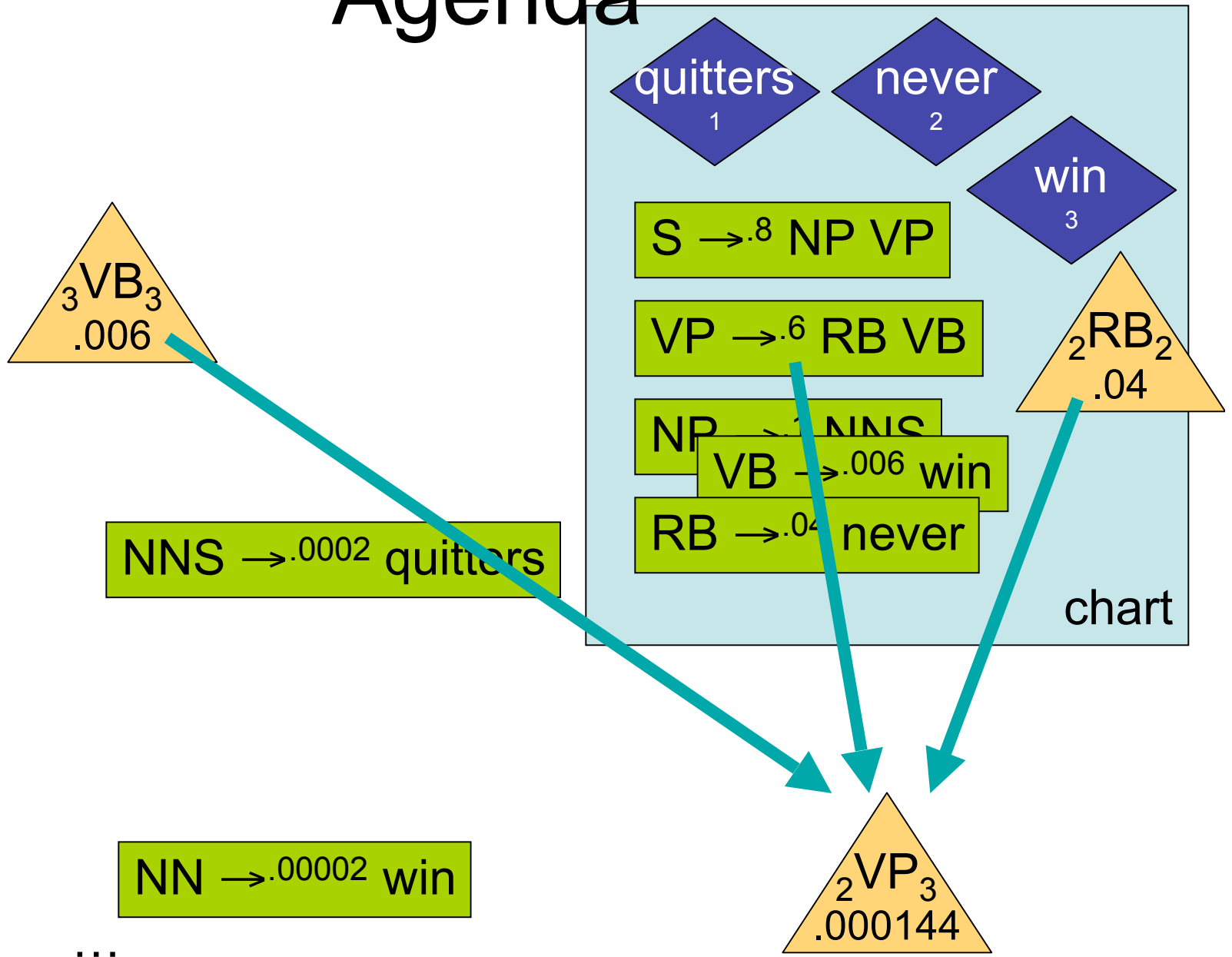


NN → .00002 win

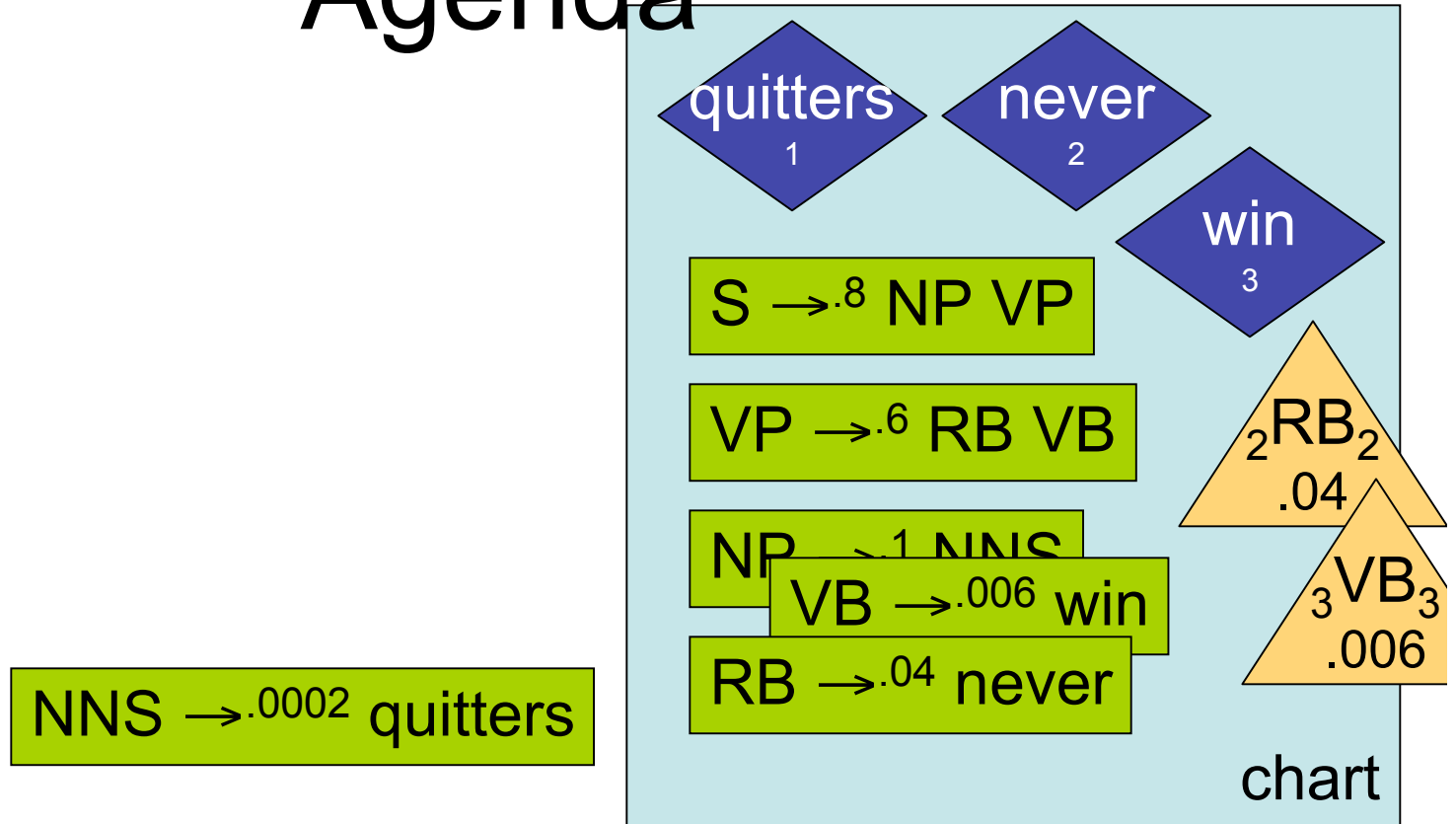
3VB₃ .006

...

Agenda



Agenda



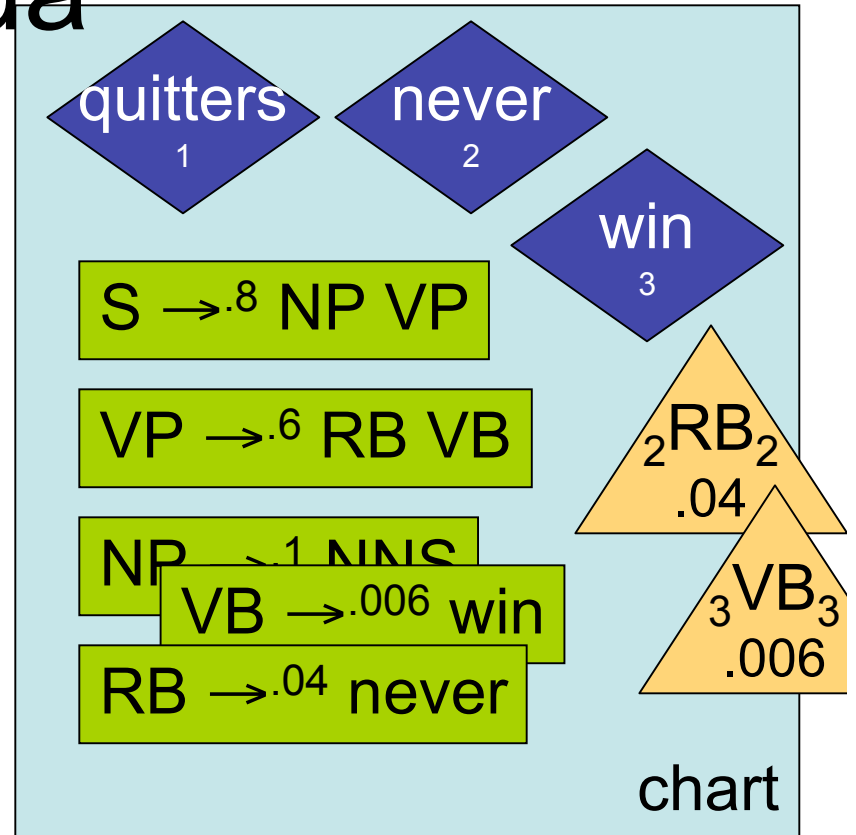
NN → .00002 win

...

2VP₃ .000144

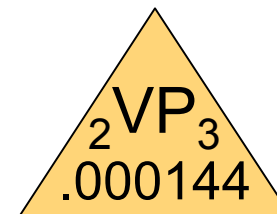
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NNS → .0002 quitters

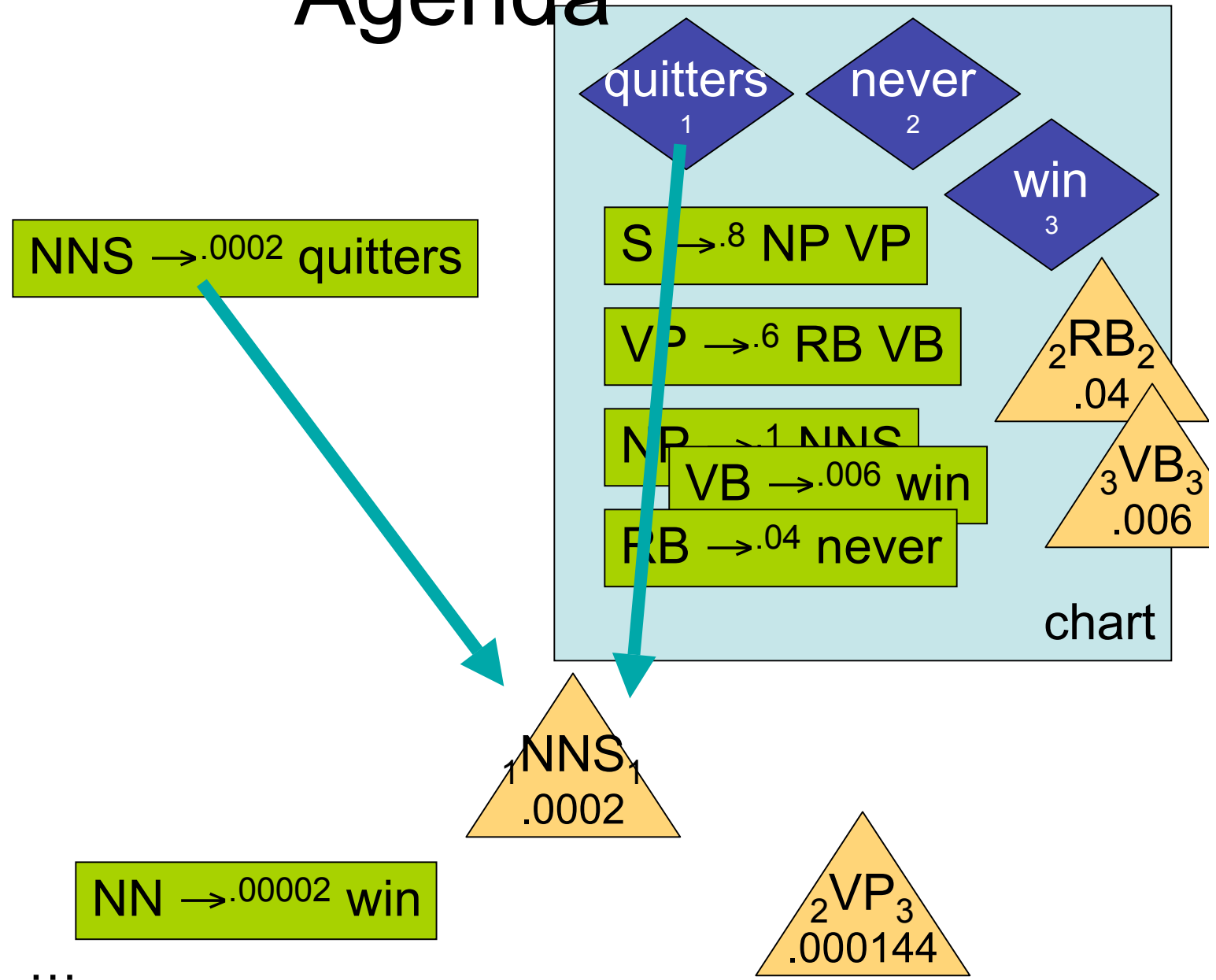


NN → .00002 win

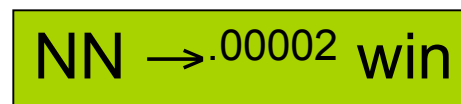
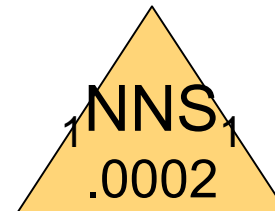
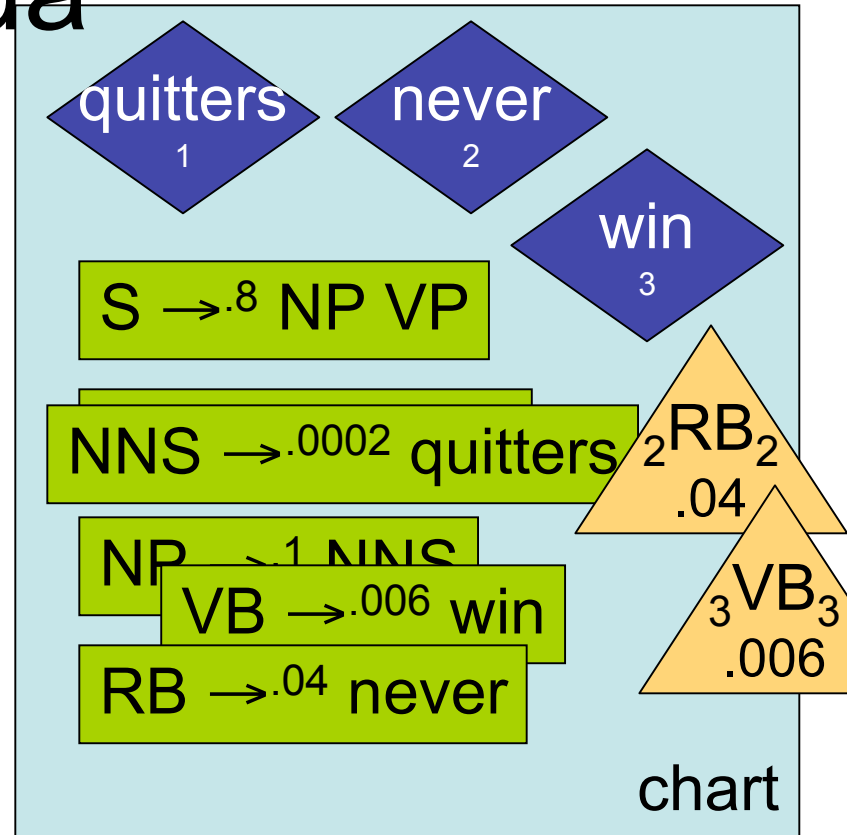
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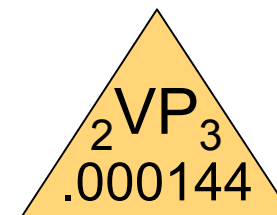
Agenda



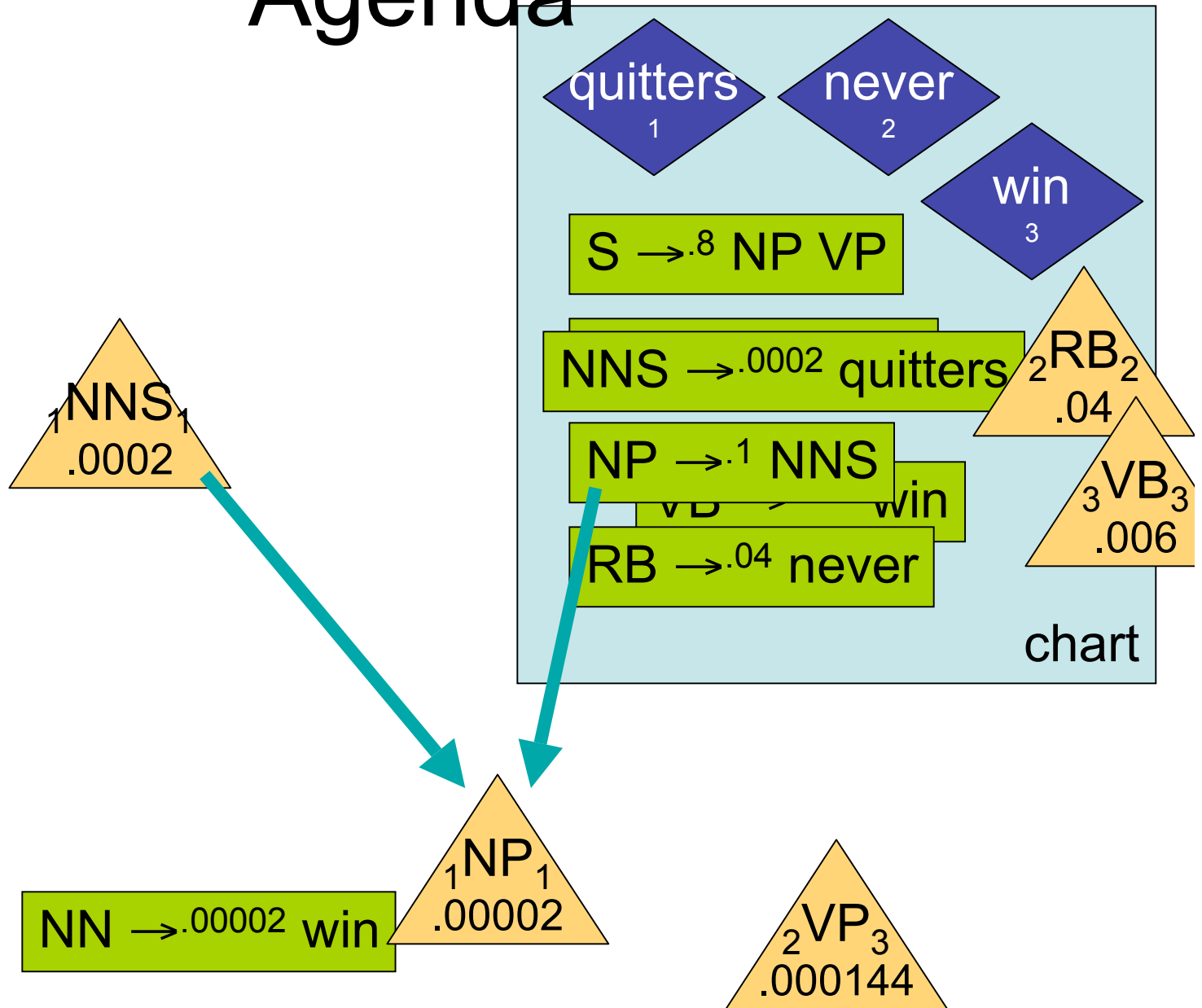
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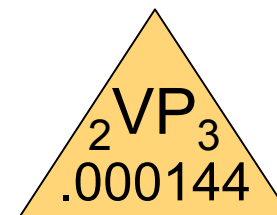
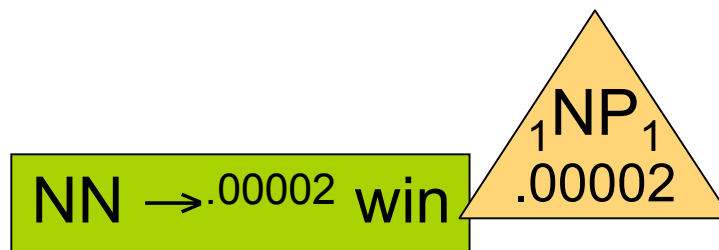
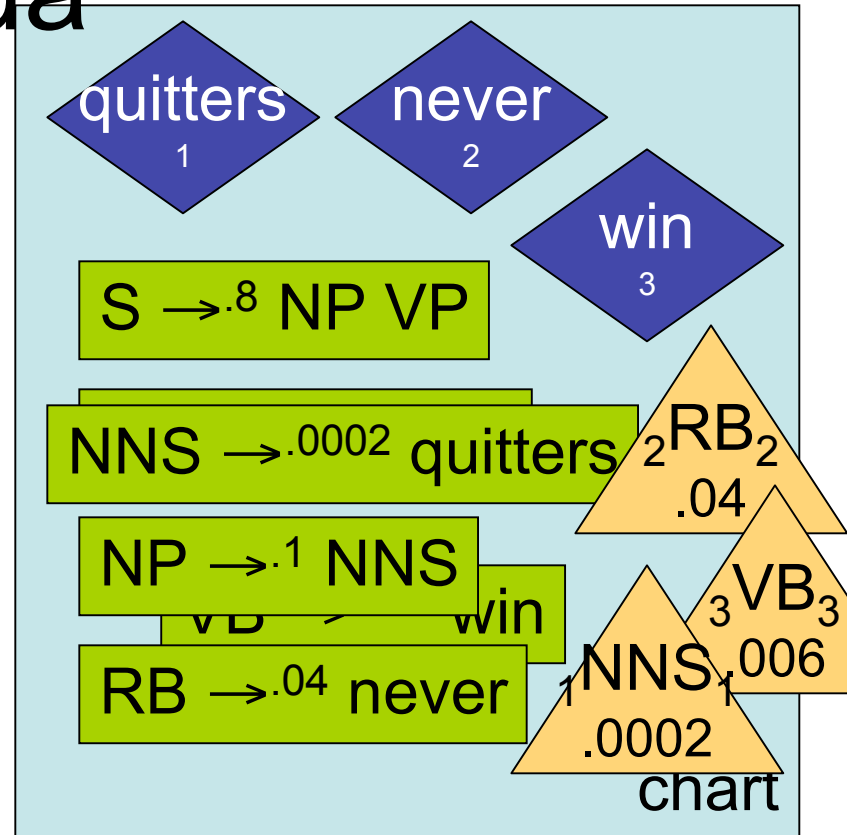
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Agenda

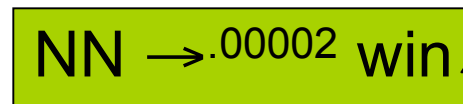
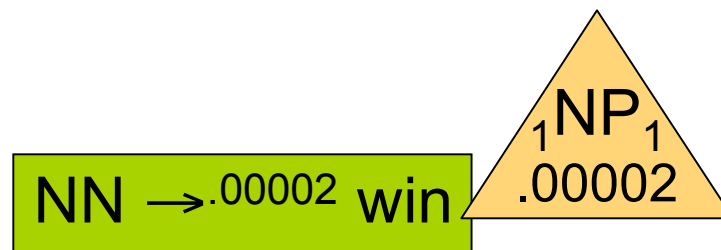
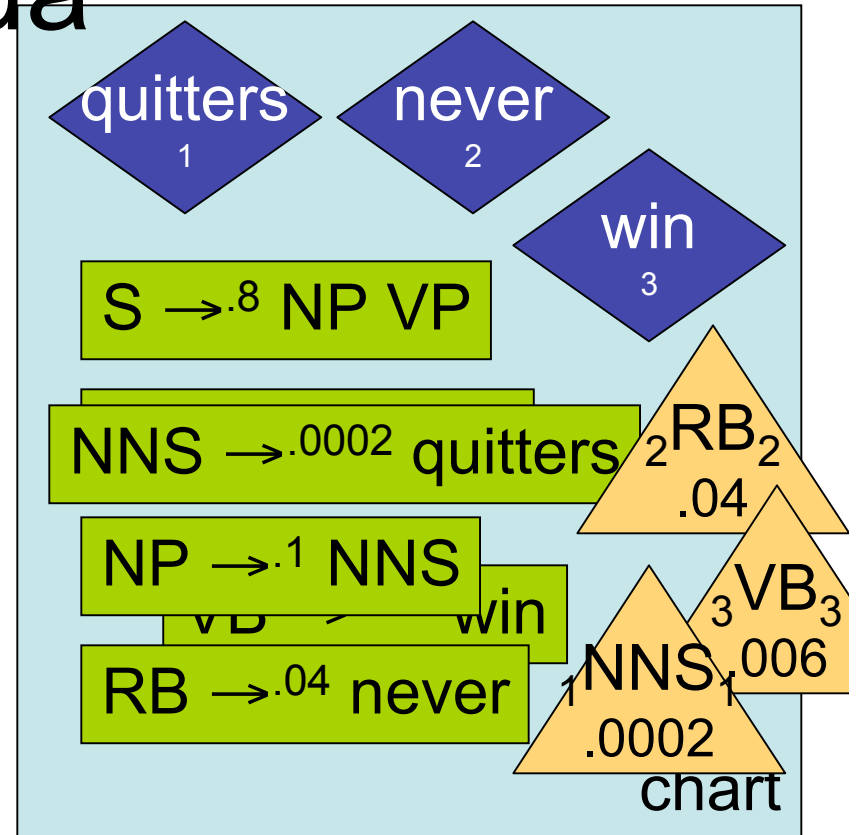
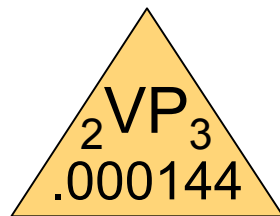


Agenda



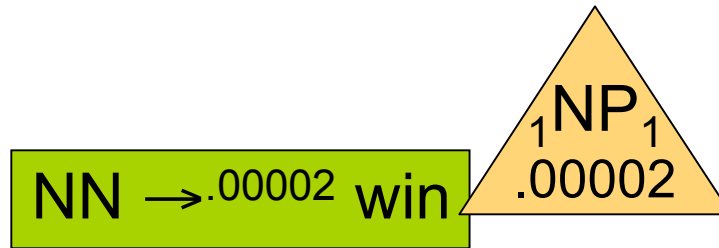
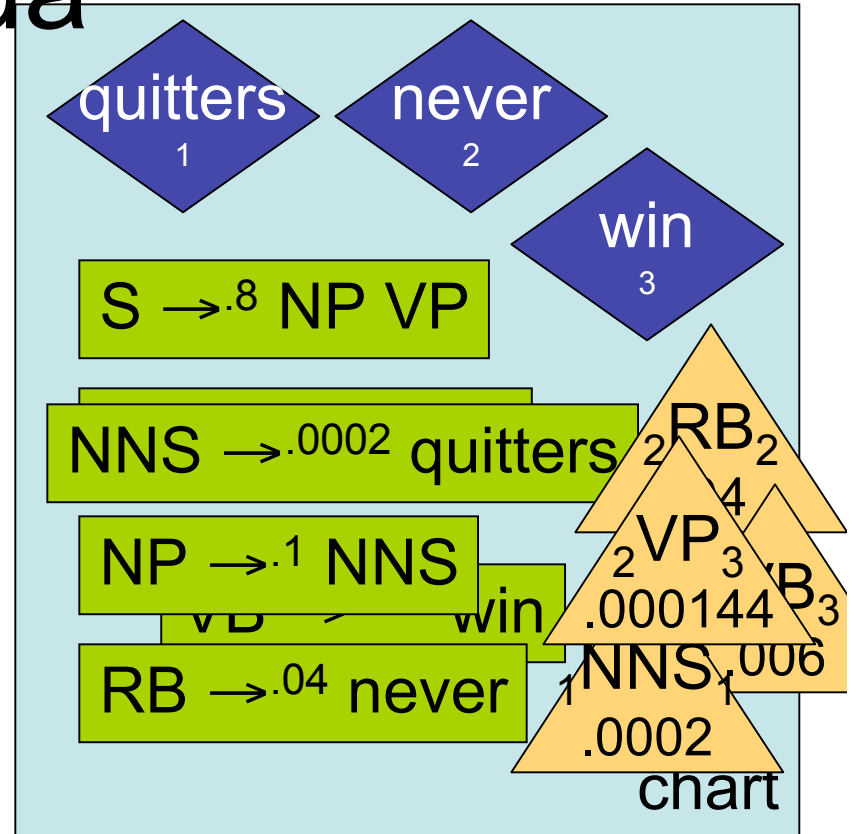
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Agenda



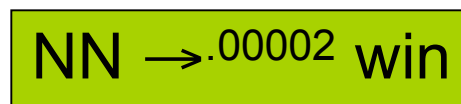
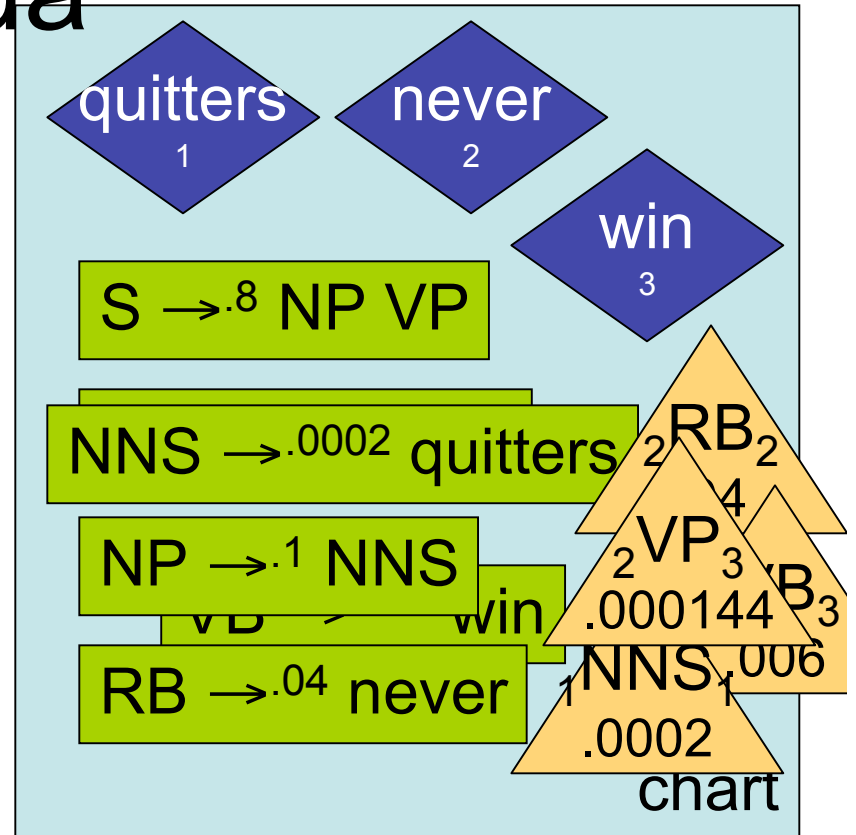
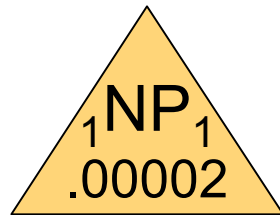
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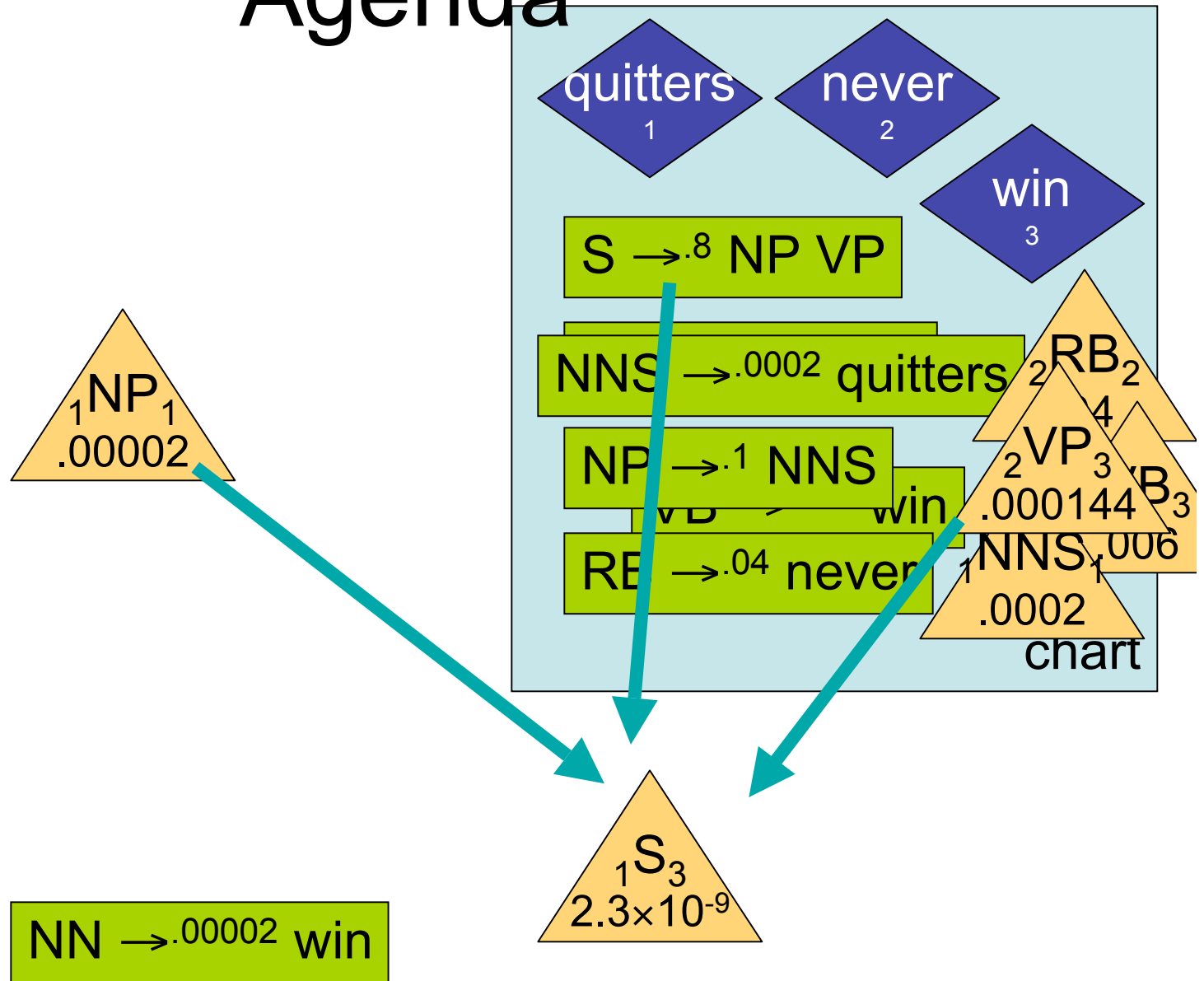
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Agenda



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Agenda



Best-First Parsing

- **Viterbi** semiring (find the best parse)
- Cf. Goodman, build the chart and fill in weights at the same time.
- Order items by their weights.
 - “**Uniform cost search**”
 - Guarantee: the first time **goal** is popped, you have the optimal parse.
- Charniak et al., 1998: heuristics to speed this up. “Figures of Merit” (big speed payoff).
- Klein and Manning (2003): admissible heuristics to guarantee optimal parse is found (big speed payoff).

Dyna

- **Dyna** is a high-level programming language (like Prolog) for weighted deduction.
- Source code looks like Prolog.
- Compiles into C++.
- Core algorithms:
 - Generalized weighted, **prioritized** agenda.
 - Allows the use of heuristics, including A^*
 - Efficient “tape” mechanism for reverse computation.
 - Very similar to backpropagation.

Dyna Programs

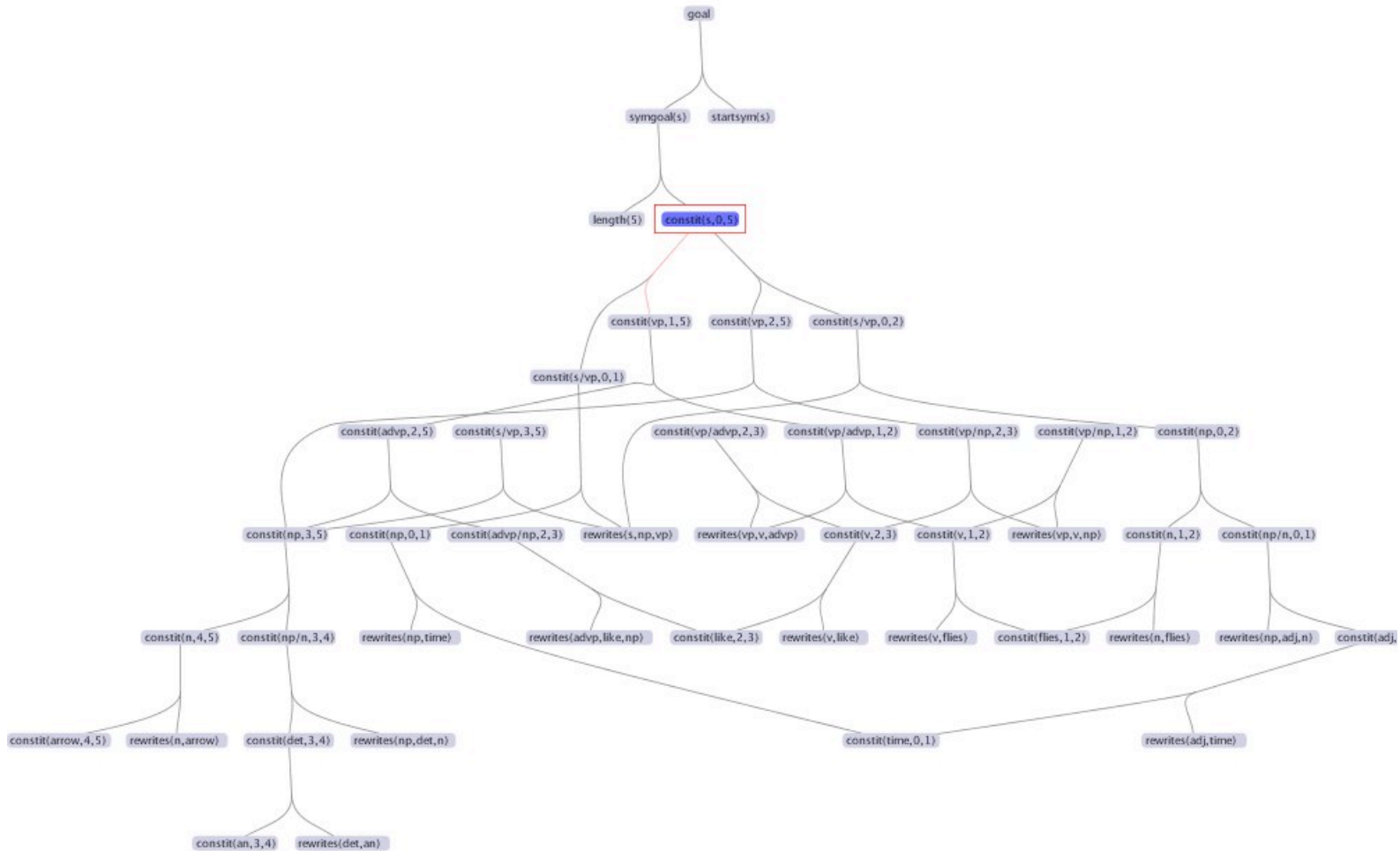
```
constit(X,I,J) += word(W,I,J) * rewrite(X,W).
constit(X,I,J) += constit(Y,I,Mid) * constit(Z,Mid,J) * rewrite(X,Y,Z).
goal           += constit("s",0,N) whenever length(N).
```

```
constit(X,I,J) max= word(W,I,J) * rewrite(X,W).
constit(X,I,J) max= constit(Y,I,Mid) * constit(Z,Mid,J) * rewrite(X,Y,Z).
goal           max= constit("s",0,N) whenever length(N).
```

```
constit(X,I,J) max= word(W,I,J) * rewrite(X,W).
constit(X,I,J) max= constit(Y,I,Mid) * inter(X, Z, Mid, J).
inter(X, Z, Mid, J) max= constit(Z,Mid,J) * rewrite(X,Y,Z).
goal           max= constit("s",0,N) whenever length(N).
```

Dyna Debugger

File Display Selection Preferences



const(s,0,5)

Shift - select nodes
LeftButton - scroll/select

Alt - stop at edges
RightButton - zoom

Ctrl - move pink trail

Parting Shots

- Weighted deduction as a convenient way to design, improve, understand, analyze, unify, and implement otherwise tricky dynamic programming algorithms.