

## From well-formed substring tables to active charts

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## Overview

- CKY algorithm:
  - explores all analyses in parallel
  - bottom-up
  - stores complete subresults
- desiderata:
  - add top-down guidance (to only use rules derivable from start-symbol), but avoid left-recursion problem of top-down parsing
  - store partial analyses (useful for rules right-hand sides longer than 2)
- Idea: also store partial results, so that the chart contains
  - passive items: complete results
  - active items: partial results

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## Representing active chart items

- well-formed substring entry:  
chart( $i, j, A$ ): from  $i$  to  $j$  there is a constituent of category  $A$
- More elaborate data structure needed to store partial results:
  - rule considered + how far processing has succeeded
  - dotted rule:  

$${}_i[A \rightarrow \alpha \bullet_j \beta] \quad \text{with } A \in N \text{ and } \alpha, \beta \in (\Sigma \cup N)^*$$
- active chart entry:  
chart( $i, j, \text{state}(A, \beta)$ )      Note that  $\alpha$  is not represented.

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## Dotted rule examples

- A dotted rule represents a state in processing a rule.
- Each dotted rule is a hypothesis:

$vp \rightarrow \bullet v\text{-ditr } np \text{ } pp\text{-to}$	We found a $vp$ if we still find
$vp \rightarrow v\text{-ditr } \bullet np \text{ } pp\text{-to}$	a $v\text{-ditr}$ , a $np$ , and a $pp\text{-to}$
$vp \rightarrow v\text{-ditr } np \bullet pp\text{-to}$	a $np$ and a $pp\text{-to}$
$vp \rightarrow v\text{-ditr } np \text{ } pp\text{-to} \bullet$	a $pp\text{-to}$
	nothing

The first three are examples of **active items** (or **active edges**)  
The last one is a **passive item/edge**.

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## The three actions in Earley's algorithm

In  ${}_i[A \rightarrow \alpha \bullet_j B \beta]$  we call  $B$  the *active constituent*.

- **Prediction:** Search all rules realizing the active constituent.
- **Scanning:** Scan over each word in the input string.
- **Completion:** Combine an active edge with each passive edge covering its active constituent.

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## A closer look at the three actions

**Prediction:** for each  ${}_i[A \rightarrow \alpha \bullet_j B \beta]$  in chart  
for each  $B \rightarrow \gamma$  in rules  
add  ${}_j[B \rightarrow \bullet_j \gamma]$  to chart

**Scanning:** let  $w_1 \dots w_j \dots w_n$  be the input string  
for each  ${}_i[A \rightarrow \alpha \bullet_{j-1} w_j \beta]$  in chart  
add  ${}_i[A \rightarrow \alpha w_j \bullet_j \beta]$  to chart

**Completion (fundamental rule of chart parsing):**

for each  ${}_i[A \rightarrow \alpha \bullet_k B \beta]$  and  ${}_k[B \rightarrow \gamma \bullet_j]$  in chart  
add  ${}_i[A \rightarrow \alpha B \bullet_j \beta]$  to chart

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## Eliminating scanning

**Scanning:** for each  $i[A \rightarrow \alpha \bullet_{j-1} w_j \beta]$  in chart  
add  $i[A \rightarrow \alpha w_j \bullet_j \beta]$  to chart

**Completion:** for each  $i[A \rightarrow \alpha \bullet_k B \beta]$  and  $k[B \rightarrow \gamma \bullet_j]$  in chart  
add  $i[A \rightarrow \alpha B \bullet_j \beta]$  to chart

**Observation:** Scanning = completion + words as passive edges. One can thus simplify scanning to adding a passive edge for each word:

for each  $w_j$  in  $w_1 \dots w_n$   
add  $_{j-1}[w_j \rightarrow \bullet_j]$  to chart

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## Earley's algorithm without scanning

### General setup:

apply prediction and completion to every item added to chart

**Start:** add  $_0[start \rightarrow \bullet_0 s]$  to chart

for each  $w_j$  in  $w_1 \dots w_n$   
add  $_{j-1}[w_j \rightarrow \bullet_j]$  to chart

**Success state:**  $_0[start \rightarrow s \bullet_n]$

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## A tiny example grammar

Lexicon:

vp  $\rightarrow$  left

det  $\rightarrow$  the

n  $\rightarrow$  boy

n  $\rightarrow$  girl

Syntactic rules:

s  $\rightarrow$  np vp

np  $\rightarrow$  det n

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## An example run

start	1. $_0[start \rightarrow \bullet_0 s]$
predict from 1	2. $_0[s \rightarrow \bullet_0 np vp]$
predict from 2	3. $_0[np \rightarrow \bullet_0 det n]$
predict from 3	4. $_0[det \rightarrow \bullet_0 the]$
scan "the"	5. $_0[the \rightarrow \bullet_1]$
complete 4 with 5	6. $_0[det \rightarrow the \bullet_1]$
complete 3 with 6	7. $_0[np \rightarrow det \bullet_1 n]$
predict from 7	8. $_1[n \rightarrow \bullet_1 boy]$
predict from 7	9. $_1[n \rightarrow \bullet_1 girl]$
scan "boy"	10. $_1[boy \rightarrow \bullet_2]$
complete 8 with 10	11. $_1[n \rightarrow boy \bullet_2]$
complete 7 with 11	12. $_0[np \rightarrow det n \bullet_2]$
complete 2 with 12	13. $_0[s \rightarrow np \bullet_2 vp]$
predict from 13	14. $_2[vp \rightarrow \bullet_2 left]$
scan "left"	15. $_2[left \rightarrow \bullet_3]$
complete 14 with 15	16. $_2[vp \rightarrow left \bullet_3]$
complete 13 with 16	17. $_0[s \rightarrow np vp \bullet_3]$
complete 1 with 17	18. $_0[start \rightarrow s \bullet_3]$

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## The Earley algorithm in Prolog

(parser/earley/earley.pl)

```
:- dynamic chart/3.           % chart(From,To,state(Lhs,Rest_Rhs))
:- op(1200,xfx,'---->').    % operator for grammar rules
```

```
% recognize(+WordList,+Startsymbol): Earley recognizer toplevel
```

```
recognize(String,Startsymbol) :-
    retractall(chart(_,_,_)),
    enter_edge(0,0,state('S',[Startsymbol])),
    scan(String,0,N),
    chart(0,N,state('S',[])).
```

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```
% enter_edge(+FromIndex,+ToIndex,+Contents)
```

```
% a) only add if it does not yet exist:
```

```
enter_edge(I,J,State) :-
    chart(I,J,State),
    !.
```

```
% b) add to chart and make try prediction/completion
```

```
enter_edge(I,J,State) :-
    assertz(chart(I,J,State)),
    predict(I,J,State),
    complete(I,J,State).
```

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```

predict(_,J,State) :-
    State = state(_,[B|_]),      % active edge
    (B ---> Gamma),
    enter_edge(J,J,state(B,Gamma)),
    fail
; true.

% -----

complete(K,J,State) :-
    State = state(B,[]),        % passive edge
    chart(I,K,state(A,[B|Beta])),
    enter_edge(I,J,state(A,Beta)),
    fail
; true.

```

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```

scan([],N,N).
scan([W|Ws],JminOne,N) :-
    J is JminOne+1,
    enter_edge(JminOne,J,state(W,[])),
    scan(Ws,J,N).

```

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### The tiny example grammar (parser/earley/earley\_grammar.pl)

```

% lexicon:
vp ---> [left].
det ---> [the].
n ---> [boy].
n ---> [girl].

% syntactic rules:
s ---> [np, vp].
np ---> [det, n].

```

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### The example run in Prolog

(parser parser/earley/earley\_trace.pl, grammar: parser/earley/earley\_grammar.pl)

```

| ?- recognize([the,boy,left]).
START:          1: 0-state(S,[s])-----0
PRED s in 1:    2: 0-state(s,[np,vp])----0
PRED np in 2:   3: 0-state(np,[det,n])---0
PRED det in 3:  4: 0-state(det,[the])----0
SCAN 1 (the):   5: 0-state(the,[])-----1
COMP 4 + 5:     6: 0-state(det,[])-----1
COMP 3 + 6:     7: 0-state(np,[n])-----1
PRED n in 7:    8: 1-state(n,[boy])-----1
PRED n in 7:    9: 1-state(n,[girl])-----1
SCAN 2 (boy):  10: 1-state(boy,[])-----2
COMP 8 + 10:    11: 1-state(n,[])-----2
COMP 7 + 11:    12: 0-state(np,[])-----2
COMP 2 + 12:    13: 0-state(s,[vp])-----2
PRED vp in 13: 14: 2-state(vp,[left])----2
SCAN 3 (left): 15: 2-state(left,[])-----3
COMP 14 + 15:   16: 2-state(vp,[])-----3
COMP 13 + 16:   17: 0-state(s,[])-----3
COMP 1 + 17:    18: 0-state(S,[])-----3
SUCCESS: 18

```

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### Improving the efficiency of lexical access

- In the setup just described
  - words are stored as passive items so that
  - prediction is used for preterminal categories. The set of predicted words for a preterminal can be huge.
- If each word in the grammar is introduced by a preterminal rule *cat* → *word* one can add a **passive item for each preterminal category** which can dominate the word instead of for the word itself.
- What needs to be done:
  - syntactically distinguish syntactic rules (--->/2) from rules with preterminals on the left-hand side, i.e. lexical entries (lex/2).
  - modify scanning to take lexical entries into account

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### Code change for preterminals as passive edges

(parser/earley/preterminals/earley.pl)

```

scan([W|Ws],JminOne,N) :-
    J is JminOne+1,
    enter_edge(JminOne,J,state(W,[])),
    scan(Ws,J,N).

```

is changed to

```

scan([W|Ws],JminOne,N) :-
    J is JminOne+1,
    ( lex(Cat,W),
      enter_edge(JminOne,J,state(Cat,[])),
      fail
    ; scan(Ws,J,N)).

```

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### The tiny example grammar in the modified format (parser/earley/preterminals/grammar1.pl)

```
% lexicon:
lex(vp,left).
lex(det,the).
lex(n,boy).
lex(n,girl).

% syntactic rules:
s ---> [np, vp].
np ---> [det, n].
```

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### The improved example run

(parser parser/earley/preterminals/earley\_trace.pl, grammar: parser/earley/preterminals/grammar1.pl)

```
| ?- recognize([the,boy,left],s).
START:          1: 0--state(S,[s])-----0
PRED s in 1:    2: 0--state(s,[np,vp])---0
PRED np in 2:   3: 0--state(np,[det,n])--0
SCAN 1 (the):   4: 0--state(det,[])-----1
COMP 3 + 4:     5: 0--state(np,[n])-----1
SCAN 2 (boy):   6: 1--state(n,[])-----2
COMP 5 + 6:     7: 0--state(np,[])-----2
COMP 2 + 7:     8: 0--state(s,[vp])-----2
SCAN 3 (left):  9: 2--state(vp,[])-----3
COMP 8 + 9:    10: 0--state(s,[])-----3
COMP 1 + 10:   11: 0--state(S,[])-----3
SUCCESS: 11
```

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### Towards more flexible control

The algorithms, we saw

- use the Prolog database to store the chart and
- Prolog backtracking on edges in chart instead of an explicit agenda.

Alternatively, one can

- explicitly introduce an **agenda**
- to store and work off edges in any order one likes.

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### Earley-recognizer with explicit agenda and chart (parser/earley/agenda/earley.pl)

```
:- op(1200,xfx,'--->'). % Operator for grammar rules
```

```
% Data structures: chart(From,To,Category)
```

```
% -----
```

```
% recognize(+WordList)
```

```
% top-level predicate for Earley recognizer
```

```
recognize(String,Startsymbol) :-
    StartAgenda=[chart(0,0,state('S',[Startsymbol]))],
    process_agenda(StartAgenda,[],Chart0),
    scan(String,0,N,Chart0,Chart),
    element(chart(0,N,state('S',[])),Chart).
```

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```
% process_agenda(+Agenda,+ChartIn,-ChartOut)
```

```
process_agenda([],X,X).
process_agenda([Edge|Agenda0],Chart0,Chart) :-
    element(Edge,Chart0), !,
    process_agenda(Agenda0,Chart0,Chart).
process_agenda([Edge|Agenda0],Chart0,Chart) :-
    Chart1=[Edge|Chart0],
    %
    predict(Edge,PAgenda),
    append(PAgenda,Agenda0,Agenda1),
    %
    complete(Edge,Chart1,CAgenda),
    append(CAgenda,Agenda1,NewAgenda),
    process_agenda(NewAgenda,Chart1,Chart).
```

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```
scan([],N,N,Chart,Chart).
scan([W|Ws],JminOne,N,Chart0,Chart) :-
    J is JminOne+1,
    setof(chart(JminOne,J,state(Cat,[])),
        lex(Cat,W),
        Agenda),
    process_agenda(Agenda,Chart0,Chart1),
    scan(Ws,J,N,Chart1,Chart).
```

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```

predict(chart(_,J,state(_, [B|_])),Agenda) :-
    setof(chart(J,J,state(B,Gamma)),
        (B ---> Gamma),
        Agenda), !.
predict(_, []). % is passive edge or no matching grammar rule

complete(chart(K,J,state(B, [])),Chart,Agenda) :-
    setof(chart(I,J,state(A,Beta)),
        element(chart(I,K,state(A, [B|Beta])), Chart),
        Agenda), !.
complete(_,_, []). % is active edge or no matching chart edge

```

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```

% -----
% element(?Element,+List)

element(X, [X|_]).
element(X, [_|L]) :-
    element(X,L).

% -----
% append(+List,?List,-List) or append(-List,?List,+List)

append([],L,L).
append([H|T],L, [H|R]) :-
    append(T,L,R).

```

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